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CONTINUOUS POROSITY CHARACTERIZATION:

METRIC-SCALE INTERVALS IN HETEROGENEOUS SEDIMENTARY

ROCKS USING MEDICAL CT-SCANNER

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Abstract

Although computed tomography (CT-Scanning) has been regularly applied to core analyses in petroleum geology, there is still a need to improve our ways to document porosity and porosity distribution in the entire pore scale spectrum, from the tens of nanometer to the meter-scale. Porosity imaging is particularly crucial for complex and heterogeneous rocks such as hydrothermally altered and fractured carbonates. The present work proposes a improved method using medical-CT to reliably estimate reservoir porosity. An in-house core-flooding setup allowed to analyse several individual core samples, scanned simultaneously (dry and saturated), as well as continuous core sections up to 1.5 m long. Without any prior knowledge of samples, three-dimensional alignment and subtraction of the two data sets (dry and saturated states) results in the generation of 3D porosity matrices. The methodology tested on a large set of reference core material shows a strong correlation between conventional gas porosimetry techniques and porosity from CT-scan. The added value of the porosity measurements by CT-scan is, first of all, the generation of 3D images of pore network, allowing to assess spatial attributes of macropores, their distribution and connectivity. Secondly, the CT-scan method also provides continuous porosity profile at the millimetric scale. Both developments are crucial for the understanding of reservoir rock properties.

Keywords: 3D porosity, carbonate, core-flooding, CT scan

INTRODUCTION

The spatial distribution of petrophysical properties of reservoir rocks commonly needs to be assessed whether it is for the exploration of hydrocarbons, the identification of efficient reservoir for CO₂ storage or the assessment of geothermal resources (e.g. Kukkonen and Peltoniemi, 1998; Mees et al., 2003; Chadwick et al., 2004; Hartmann et al., 2008; Shi et al., 2009; Goldberg et al., 2010; Perrin and Benson, 2010a). Computed tomography (CT-scanning) has been applied for decades by the hydrocarbon community to supplement conventional core analyses (such as mineralogy, porosity, permeability) or wireline logging interpretation (Vinegar and Wellington, 1987; Wellington and Vinegar, 1987; Dulu, 1999; Akin and Kavscek, 2003; Taud et al., 2005; Geiger et al., 2009; Baniak et al., 2013). CT-scanning techniques are non-destructive and offer the possibility to quantify internal structures based on the measurement of X-Rays attenuation coefficients, which depend on the chemical composition and physical density of the materials analysed (Dulu, 1999; Akin and Kavscek, 2003; Cnudde and Boone, 2013). Therefore, CT-scanning provides qualitative analysis when applied to subsurface materials (e.g. heterogeneity, damages, presence of fluids); it can also be used to gain quantitative information on cores such as bulk density, porosity, and fluid saturations (Coles et al., 1991; Cnudde and Boone, 2013). Efforts have also been made to use CT-scanning to understand fluid displacement and relative permeability of core material from reservoirs (Hove et al., 1987; Vinegar and Wellington, 1987; Withjack, 1988), notably because of their importance in oil-recovery processes during production stages (Andrianov et al., 2012; Simjoo et al., 2013; Simjoo and Zitha, 2018). Limitations of medical CT (or conventional CT, as proposed by Ketcham and Carlson (2001)) are well

known. The image spatial resolution obtained is relatively low (about 0.5 mm on average), allowing characterization and quantification of the spatial distribution of larger structures only such as burrows, roots, primary or secondary framework pores or fractures. The use of medical CT systems is consequently being complemented by MicroCT and synchrotron based systems (Ketcham and Carlson, 2001; Cnudde and Boone, 2013; Wildenschild and Sheppard, 2013) that offers a better resolution.

However, there is a critical need to get data on the entire pore scale spectrum, from the tens of nanometer to the meter-scale in order to build comprehensive datasets for rocks, sediments and soils (Okabe and Blunt, 2007; Biswal et al., 2009; Vaz et al., 2014; Bultreys et al., 2016b; Xiong et al., 2016). Porosity imaging and pore-network modelling techniques are crucial for characterizing complex and heterogeneous rocks such as carbonates (Sok et al., 2010; Pak et al., 2016). Deposition processes and diagenesis result into heterogeneity at multiple spatial scales so that predicting petrophysical properties is particularly challenging in carbonate reservoir rocks. Pore structure affects connectivity, conductivity and permeability that all influence oil recovery mechanisms, a key aspect for economic geology. MicroCT and synchrotron-based systems give access to spatial resolutions of few micrometers to deal with pore-particles interfaces but sample size is restricted to few millimeters (Vaz et al., 2014) and is time-consuming. The trade-off between sample size and spatial resolution (i.e. voxel size), mainly imposed by the micro-CT scanner detector characteristics, is therefore a limitation to upscale data obtained at the pore scale and justifies modelling. Even if significant developments are made regarding small scale heterogeneity characterization techniques, there is a need to

optimize characterisation of heterogeneous material such as fractured or dolomitized carbonates at the centimeter scale.

Coupled with core-flooding experiments, CT scanning are well establish in reservoir studies to monitor the progress of multiphase displacement inside porous media (Schembre and Kovscek, 2003; Shi et al., 2009). Different core-flooding setups exist, depending on the fluids involved or the experimental conditions, such as the confining pressure that can be radial and/or axial, the temperature (reservoir condition), or the fluid injection mode (Wang et al., 1984; Alemu et al., 2013). The reliability of such scanning technique was verified through a study of oil and gas shows in eastern Québec, where an industry, university and government partnership was developed to improve current CT scan methods, which remained difficult to use when characterizing reservoir properties in heterogeneous carbonates. The authors found, to the extent of their knowledge, that rigorous and complete scientific literature on CT scan technique was needed to implement a detailed characterization program with specific interest to carbonate rocks. Studied rock units were locally found within highly fractured intervals associated with replacive and pore-filing hydrothermal dolomites. The industry partner, Squatex Inc., drilled and cored over 6000 m of stratigraphic exploration wells and needed an efficient and rapid way to assess the spatial distribution of reservoir properties in the area in order to plan their future works.

This project objective was to complement and enhance current CT scan methods applied to sedimentary rocks in order to reliably document porosity distribution and connectivity in heterogeneous carbonate reservoirs and to predict its distribution over a metric scale. An improved methodology combining a core-flooding setup and medical-CT analyses to

obtain the porosity of heterogeneous rock material was consequently developed. No confinement pressure is applied during saturation: water flows around sample and the experiment is conducted at room temperature. The methodology has been tested on reference core material that has a wide range of porosity values and includes eight different sedimentary rock types. To test this enhanced CT methodology, porosity results are compared with conventional ways to determine porosity in rock material (i.e. helium gas porosimetry). In addition, the methodology was tested on a specific metric section of an heterogeneous carbonate interval. Within this interval, 3D porosity matrices were generated to visualize the connected porosity network, as well as a continuous porosity profiles at the millimetric scale.

PREVIOUS WORKS

1) CT-scanning

Both medical and microfocus CT-scanning (Micro-CT) has been commonly used for decades for core analyses in the Oil and Gas sector in order to analyse porosity, fractures patterns, or assess fluid flow in porous rocks (e.g. Honarpour et al., 1985; Grader et al., 2000; Van Geet and Swennen, 2001; Akin and Kovscek, 2003; Karacan et al., 2003; Van Geet et al., 2003; Denney, 2004; Taud et al., 2005; Geiger et al., 2009). Images generated contains relative density measurements in Hounsfield Unit (HU) and allow qualitative, such as internal structures description, and quantitative information analysis. Different parameters can be derived from HU values as it relates to three physical aspects: real density, chemical composition (atomic number) and porosity. To isolate each properties, different methods has been developed. The atomic number can be computed using Dual-energy scans (Wellington and Vinegar, 1987; Alves et al., 2014; Jussiani and Appoloni,

2015). Density estimation in g/cm^3 is often estimated using calibration points obtained from core plugs. Finally, porosity is estimated using various approach: (1) segmentation of pores, (2) mixels (or mixed pixels(Kato et al., 2013) assuming grain density and (3) saturation technique and subtraction of saturated and dry state (Withjack, 1988; Davis et al., 1992).

2) Core-flooding system

From the 1950's, hydrocarbon exploration companies have commonly used X-rays to study reservoir properties (Morgan et al., 1950) and, with the advance of 3D computed tomography, to visualize fluid flow through reservoir rocks, calculate porosity and oil saturation (Vinegar and Wellington, 1987; Wellington and Vinegar, 1987; Withjack, 1988). Numerous papers provided excellent syntheses on computed tomography principles and common practices (e.g. Newton and Potts, 1981; Ketcham and Carlson, 2001). With respect to core flooding experiments, tests under CT are primarily made to observe fluids displacement and to gain information about relative permeabilities (e.g. Wang et al., 1984; Hove et al., 1987; Wellington and Vinegar, 1987; Soltani et al., 2009), either comparing two liquid phases (e.g. oil, brines) or using a gas phase (e.g. CO_2) and a denser liquid phase (Schembre and Kovscek, 2003; Shi et al., 2009; Perrin and Benson, 2010b; Alemu et al., 2013; Krause et al., 2013; Krause and Benson, 2015; Jackson et al., 2018). In more recent years, core-flooding studies were carried out with one sample at the time only and were usually associated with a dynamic set-up, using pump(s) to ensure constant fluid circulation through the core holder and samples. Most common setup involves a water filled chamber made of multi layers sleeve and hosting the sample. The chamber is maintained at particular confining pressure and temperature

in order to reproduce reservoir conditions at depth (Perrin and Benson, 2010a; Krevor et al., 2012; Pini et al., 2012).

In the late 1980's, Withjack (1988) presented a protocol to measure saturation and porosity with a core-flooding device under CT. This work was based on X-rays and porosity determination principles recognized years before (Morgan et al., 1950; Laird and Putnam, 1959). The setup of Withjack (1988) first involved core samples placed in an aluminium chamber and scanned in a dry state. The aluminum chamber was used to remove lower-energy photons and limit beam hardening artefacts. Prior to the experiment, polyethylene bottles filled with sodium iodide solutions were scanned separately inside the core holder. Then core samples were removed from the chamber, put under vacuum conditions and immersed in the sodium iodide solution. Once samples were considered saturated, they were put back in the chamber and scanned under CT. Withjack (1988) tested a Berea sandstone and a dolomite with both porosity around 20%. Porosity was also measured by the re-saturation method for comparison purposes. Although CT-scan capabilities remained limited at that time, Withjack (1988) demonstrated that CT-scan porosity values correlated well with those obtained by re-saturation ($\pm 1\%$). The approach described in this paper is primarily based on the so-called X-ray saturation technique developed by Withjack (1988), but includes recent advances such as procedures for registration and to reduce image noise (e.g. Ketcham and Iturrino, 2005; Pini et al., 2012; Pini and Madonna, 2016).

3) Doping agents used with CT-scan

A doping agent is a fluid with dissolved salts having a high atomic number ($Z \geq 50$), which alters X-rays absorption properties of the fluid phase. This enhances the contrast

between solid and fluid phase (Wildenschild et al., 2002). The use of such dopant is common in the hydrocarbon industry when working on multiphase fluid flow experiments (e.g. Vinegar and Wellington, 1987), and the most popular choices for doping brines are sodium bromide (NaBr), sodium molybdate (Na_2MoO_4) and sodium iodide (NaI). Doping agents are also widely used in other research areas such as soils science (Hopmans et al., 1992; Anderson et al., 2003; Helliwell et al., 2013; Vaz et al., 2014).

In the frame of this paper, we present an improved methodology in a sense that we combine, as simple as it can, a core flooding system that does not require any confinement. In addition, the setup allows the saturation of multiple meters of samples simultaneously, thus minimizing the operating cost. In order to validate the methodology, we studied an extended range of porosity (~1% up to ~30%) and different sedimentary rock types, including sandstone but with specific interest to dolomite and limestone.

METHODOLOGY

1) Samples

1-1 Reference samples

Reference core materials were obtained from two distinct vendors located in the USA: Cleveland Quarries (Vermilion, Ohio) and Kocurek Industries Inc (Caldwell, Texas). Eight different lithologies commonly used as test material in the hydrocarbon industry or rock-mechanics studies were selected (e.g. Churcher et al., 1991; Devarapalli et al., 2017; Islam et al., 2018). These lithologies were chosen to cover a large range of porosity (1.5 to 28 %; Fig. 1; Tab. 1). All reference core samples are cylindrical and diameter is either

3.8 or 4.5 cm. For this work, tested variables therefore include lithology type, porosity range, core diameter and length.

The Berea sandstone is medium-grained, Mississippian in age and outcrops in Ohio (USA). It is mainly composed of quartz and feldspar with small clay content. Particles are well sorted and subangular with quartz overgrowths. Berea sandstone is probably the most commonly used rock as an analogue for hydrocarbon reservoirs (Pini and Madonna, 2016). It has a porosity of approximately 20% (Winkler, 1983; Churcher et al., 1991; Hart and Wang, 1995; Boon et al., 2017). The Nugget Sandstone is fine-grained, laminated, Jurassic in age. This sandstone outcrops in Utah and Wyoming (USA). Its porosity can reach up to 25% (Lindquist, 1988) and it is essentially composed of angular to subangular quartz grains (Fig. 2A). The Boise sandstone is a medium to coarse-grained Late Miocene sandstone (Winkler, 1983) from Idaho (USA). This sandstone is poorly sorted and composed of quartz and feldspar, with minor clay (Fig. 2B). It has a porosity of approximately 28-30%. The Scioto sandstone is homogenous, fine-grained, Mississippian in age and outcrops in Ohio (USA). Quartz grains are subangular (Figs. 2C-D). This sandstone commonly has a porosity of approximately 12% (Holder et al., 2001; Bose et al., 2014). The Indiana limestone, also known as the Salem Formation or the Bedford limestone, is a middle Mississippian marine bioclastic carbonate that crops out in south-central Indiana (USA). The limestone is heterogeneous, grainstone to packstone and is mostly composed of calcium carbonate with only little amount of magnesium carbonate. Fossil fragments include bryozoans, echinoderms, brachiopods (Churcher et al., 1991). The Indiana limestone is relatively well cemented and has a porosity of approximately 13% (Musselman, 1967; Schmidt and Huddle, 1977; Hart and Wang, 1995; Boon et al.,

2017). The Carthage marble, the commercial name for Carthage limestone or Burlington limestone, is a homogenous well cemented crinoid-rich Mississippian limestone (Fig. 2E) that outcrops in Missouri (USA). It is composed predominantly of calcium carbonate with little amount of glauconite. This limestone exhibits low porosity value, usually about 1.5% (Musselman, 1967; Martin, 1968). The Silurian Dolomite, is a homogeneous fine grained dolostone in Ohio (USA). It is composed predominantly of non planar dolomite crystals arranged in mosaic (Fig. 2F) and its porosity is usually around 20%, but could be down to 14% (e.g. Islam et al., 2018). The Guelph dolomites, also known as the Baker Dolomite is a homogeneous dolomitized carbonate sand, formed in marine shallow water environment. Even though two types of Guelph dolomites exist, the samples used in this work are fine-grained, light gray dolomite (Churcher et al., 1991). The Silurian Guelph dolomites is present in Ohio (USA) and Ontario (Canada). Its porosity value is usually around 7% but can be much higher (up to 24%).

1-2 Silurian core samples

Silurian core samples being studied for this paper come from the Gaspé Belt in Québec (sensu Bourque et al., 1995) (Fig. 3). It primarily consists of Upper Ordovician to Middle Devonian fine to coarse nearshore to deep marine clastic deposits with subordinate shallow- to deep-water carbonate platform deposits. The industry partner, Squatex Inc. has drilled several stratigraphic wells over the past ten years. Promising reservoir intervals are located in the cyclic offshore – peritidal carbonates of the lower Silurian Sayabec Formation and the underlying shoreface clastics of the lower Silurian Val-Brillant Formation (Lavoie and Bourque, 2001; Lavoie and Morin, 2004). Even if these two units are not hydrocarbon producers, both have ubiquitous evidence of at least

locally, hydrocarbon charge (Lavoie et al., 2009; Dietrich et al., 2011). Hydrothermal dolomites of the Sayabec Formation are found in the well bedded intertidal to shallow subtidal facies (Lavoie and Morin, 2004); they host an exhumed oil field a few kilometers to the east of the study area (Fig. 3) with bitumen and dead oil filling matrix porosity and fractures (Lavoie and Chi, 2010). While drilling in the Témiscouata area (Fig. 3), oil or gas shows were associated with naturally fractured hydrothermal dolomites (HTD) intervals within the Sayabec Formation as well as with the underlying sandstone in the Val-Brillant Formation. In both cases, the shows seemed to be associated with fractured intervals (pers. comm., S. Larmagnat, 2015).

Silurian samples were recovered from 1 7/8 inch (47.625 mm) diameter drilled cores. All samples were taken from a specific hydrothermal dolomite interval (Fig. 4A), corresponding to a section of 1.8 meters long. Due to the lack of integrity of some portions, only 18 cylindrical core samples with a minimum length of 5 cm have been selected. Out of this 18 samples set, five samples were sent for helium porosimetry measurements at an external lab, AGAT Laboratories, after the core-flooding experiment was completed (Fig. 4B). Those samples were selected on the basis of representability of the whole 1.8 meter succession. Hence we have selected visually non-porous limestone muds, limestone mud with some visible pore space, vuggy dolostone, fractured dolostone and vuggy and fractured dolostone to cover the entire spectrum of depositional/diagenetic elements present in those carbonates.

2) Helium porosity measurements

3-1 Reference rock samples

Because reference rock samples were not provided with exact porosity values from the core vendor itself, all reference samples were sent to the external laboratory, namely AGAT laboratories in Calgary (Canada), to get helium gas porosity measurements. AGAT Laboratories are private, independent, and routinely run petrophysical properties analyses (including helium gas porosimetry) for the oil and gas industry, academy and governmental research teams (e.g. Connell-Madore and Katsube, 2007; Black, 2014; Gasaway et al., 2018). In addition, depending on sample diameter and length, reference rock samples were analysed by two additional gas porosimetry instruments at INRS. From a total of 30 samples, ten samples have been analysed for gas porosimetry by three distinct instruments and the CT-scanner. All reference samples were analysed at least by one gas porosimeter and the CT-scanner. In all cases, gas porosity measurements were conducted on samples before analysing them under the CT-scan.

Gas porosity measurements were obtained from instruments relying on Boyle's law: the pressure exerted by a mass of helium gas is inversely proportional to the volume of the samples. Measuring the change in helium pressure gives the grain volume. Therefore, porosity value results from two measurements, grain volume and bulk volume. The first is measured by the instrument itself and the second is obtained manually using a caliper. Assuming a perfect cylindrical geometry, a caliper is used to measure the length and diameter, and calculate the sample bulk volume. For this study, the calculation of the porosities using the three different instruments have been made using a single bulk

volume for each sample in order to minimize potential discrepancies that could arise from three different manual caliper measurements.

A total of 30 samples were thus sent to AGAT Laboratories. Prior to analyses, samples are dried in a convection oven for 48 h at 108°C. The helium pressure is set at 2.76 MPa. Porosity values were given a margin of error of ± 0.005 (0.5 %). Hereafter this equipment will be referred as in-house AGAT gas porosimeter (IHAP).

A total of 20 porosity measurements were made using the Core Test Systems AP-608 Gas permeameter-porosimeter available at INRS “Laboratoire ouvert de géothermie” (LOG; Québec, Canada). Prior to analyses, samples were first dried at 108°C for at least 48 h, using a Thermolyne oven (Thermo Scientific). Initial helium pressure is set at 1.38 MPa. In order to examine the reliability of analysis results, each core plug was analysed three times and the results were averaged. This procedure delivered porosity results with an average standard deviation of 0.18. Only 20 reference samples were analysed because of diameter and length restrictions associated with the AP-608 instrument (Tab. 1-2).

Few samples were also analysed at the INRS laboratory for Decontamination and Waste Reclamation, using a gas pycnometer (Micromeritics AccuPyc 1330). Routinely used to obtain the density, the device uses the Boyle’s Law to measure grain volumes. The instrument is fully-automatic and makes three consecutive runs for each analysis. Only 10 reference samples were analysed because of diameter and length restrictions associated with this instrument (Tab. 1-2).

3-2 Silurian core samples

Silurian core samples with their 1 7/8 inch (47.625 mm) diameter could only be analysed by the IHAP at AGAT Laboratories. Out of the 18 cores from the hydrothermal dolomite interval (Fig. 4), five samples were sent to AGAT Laboratories, after the core-flooding experiment was completed. These five samples were chosen to represent the natural heterogeneity of the interval. A detailed description of all five sample is given (Figure 4B).

3) Core-flooding experiment

4-1 System

An in-house core-flooding system was designed to accommodate different sample sizes (diameter and length) and to control experimental variables. The system was built to be simple, reliable and cost effective (Fig. 5). One (or several) horizontal chamber, made of polyvinyl chloride (PVC) or acrylic (plexiglass), an X-ray-transparent material, is connected to two pumps and a water tank. The core holder inside diameter is 41.7mm and 50.8mm for 1 1/2" and 1 7/8" samples, respectively. Between chambers, Swagelok quick connects (valves) are used because they allow minimal air inclusion and minimal spillage. The laboratory vacuum pump (Welch, WOB-L Pump 2585) is used to adjust and monitor the vacuum level. A diaphragm pump (SHURflo, 2088 serie) is used to ensure constant fluid circulation through the chamber at a flow rate 19.28 l/min. Within the chamber, the internal fittings use customized 3D printed PLA (polylactic acid) core holders to increase stability and create space for water flow around the samples, particularly when the chamber is flooded. For the purpose of this work, core-flooding was performed either with distilled water or sodium iodide solution. All reference samples

(n= 30) were saturated using distilled water. In addition, 15 samples (see Tab.4) were saturated using sodium iodide salt (NaI). NaI was chosen because its behaviour is similar to NaCl with respect to argillaceous minerals (Withjack, 1988). In addition, it has a reasonable cost, approximately 170\$ per 100g, and can be handle safely without causing hazards.

4-1 Protocol

Core-flooding experiments were conducted at room temperature and no confinement pressure was applied (workflow is summarized in Fig. 6). Prior to scanning, samples were dried at 108°C for at least 48 h, using a Thermolyne oven (Thermo Scientific). In agreement with Ketcham and Iturrino (2005) , a two-stage scanning protocol was used. Samples were placed in the sealed chamber and first scanned in a dry state. Secondly, vacuum is applied for 24hours, degassing distilled water and samples simultaneously. Then the chamber is flooded with either water or NaI solution (15 g/L), during which the diaphragm pump guarantees constant fluid circulation. This removes air bubbles that could be created during the saturation process, thus maximising fluid contact and reducing the number of connected pores that would not be saturated otherwise. It also constantly provides degassed water at sample-water interface thus extracting residual air bubbles. The use of NaI doping agent, with a concentration of 15 g/L, increases contrast by 30% when compared to water. Such improvement can be crucial when investigating small pore size that falls below the range of medical CT resolution. Core samples were weighted before and after saturation with either water or NaI to evaluate the performance of the saturation process. Sample weight was measured with a Sartorius top-loading balance (CP4202S) having a 0.01 g accuracy. According to the Liquid saturation

technique (API, 1998), pore volume can be calculated using bulk volume and known fluid density (water and NaI).

4) CT-scan measurements

CT measurements were performed using a Siemens SOMATOM Definition AS+ 128 at INRS-ETE. The X-ray tube of the CT scanner was operated at a voltage of 140 kV and a current of either 700 or 350 mAs for reference samples and core samples of the study area, respectively. The same voltage for all CT acquisition was used in order to obtain comparable HU values. However, the current had to be adapted to minimize downtime due to X-ray source cooling. Longer core holder, such as the one used for the Silurian carbonate interval, allowed more samples to be processed simultaneously. The source current was then reduced from 700 mA to 350mA, thus using less power and generating less heat. Downtime between scans, using high X-ray source power, can be more than 15 minutes, adding hours of waiting time thus increasing significantly the operating cost. An H70h convolution kernel was used for the reconstruction of the images. The thickness of each CT slice was set to 0.6 mm (Tab. 3). Images were recorded in DICOM format and visualizes with the open-source software Fiji (Schindelin et al., 2012).

5) Noise reduction

Density changes associated with the infiltration of water can be subtle when porosity is low. In such cases, image noise is problematic and could outweigh the density variation associated with the water saturation. The Pini and Madonna (2016) approach was therefore adopted here to examine how the level of noise changed when averaging several scans or decreasing the resolution, and how this ultimately affected the porosity

calculation (Fig. 7). The scan repetition for this study was set to three in order to get a short acquisition time with a low image noise from this method.

6) Beam hardening

The beam hardening is a common artefact caused by the absorption of low energy photon at sample frontier thus “hardening” the beam by making the mean energy higher. This phenomena, linked with the polychromatic nature of the X-ray spectra emitted, produce density under-estimation at sample center (Ketcham and Carlson, 2001). A software beam-hardening correction to detector readings is applied by the scanner and is optimized for human body, which mainly consist of water. This correction does not remove the artefact due to rock samples. Ketcham and Itturino (2005) have showed that the sample geometry changes the beam hardening profile thus a calibration wedge with a similar density and diameter is required. The strategy to minimize the beam hardening profile variation from dry to saturated was to build a core holder with an internal diameter close to the sample diameter. This creates a nearly identical geometry between dry and saturated state, thus producing a near equivalent beam hardening profile. Moreover, the image subtraction applied at the next stage minimizes the influence of beam hardening.

7) Data analysis

Data analysis is based on the X-ray saturation technique (Withjack, 1988) but also includes recent developments (Ketcham and Iturrino, 2005; Pini et al., 2012; Pini and Madonna, 2016). Algorithms applied aim at determining porosity by comparing CT images in a saturated state and unsaturated state. The working hypothesis considers a voxel as a mixel that is a mixture of porosity and solid phase material. Density value of one mixel is therefore an average value of its content. When saturating the samples with

water or NaI, the connected pores, initially filled with air, is filled with the liquid phase. Since solid phase density and quantity do not change, recorded changes in voxels density are interpreted as the results of pore filling. Equations below (1) summarize the calculation of porosity from density matrices acquired (D).

$$\%_{porosity} = \frac{D_{dry} - D_{saturated}}{D_{gas} - D_{fluid}} \quad (1)$$

The calculation is applied for each voxel containing the sample. Grain density is not needed to evaluate porosity since only fluid density in pores changes (Boespflug et al., 1994). Fluid density is known as part of CT calibration ($D_{gas} = D_{air} = -1000$ HU; $D_{fluid} = D_{water} = 0$ HU), while the density of the NaI solution in Hounsfield unit was obtained by in situ calibration ($D_{NaI} = 324$ HU, for a 15g/L NaI solution).

The subtraction of the saturated vs dry data to calculate the porosity was performed using MATLAB®. Prior to subtraction, data registration was performed for each analysis using intensity-based image registration algorithm (MathWorks, 2018). The 3D matrix resulting of this subtraction allowed to visualize the effective pore network and the evaluation of the porosity distribution using statistics. In some instance, a circular binary mask has been used to include large vugs located at sample surface. The algorithm uses Hough transform and phase-coding (Yuen et al., 1990).

RESULTS

The results obtained in this study are divided into three main sections. The two first sections correspond to porosity measurements made on the reference samples set only ($n = 30$), either using the different helium gas porosimeters (i.e. AccuPyc, AP-608 and IHAP); or using the improved medical CT-scan methodology (Tab. 4). The third section

presents the results obtained for the silurian core samples, using both IHAP and the improved medical CT-scanning methodology (Tab. 5).

In addition to exact porosity values obtained from different instruments (AP-608, AccuPyc, IHAP or CT-scan), several correlation ratios are then considered in these results sections. Firstly, because the helium porosity measurements made using the IHAP are considered as true values, the notion of absolute error is calculated with respect to the helium porosity (Table 4-5). Second, the absolute error (AE) corresponding to the amount of error in the porosity measurements is calculated, i.e. from the difference between the porosity calculated using the improved CT-scan methodology and the porosity measured using a conventional gas porosimeter. The R-squared (R^2), commonly used in classical regression analysis (Rao et al., 1973), is calculated and represents a statistical measure of how close the data are to the fitted regression line. Also known as the coefficient of determination, R^2 ranges from 0 to 1. In the present work, R^2 is calculated to compare two instruments or methodology evaluating the porosity. Finally, the root-mean-square error (RMSE) is calculated. It corresponds to the standard deviation of the residuals (prediction errors). Residuals are a measure of how far from the regression line data points are. RMSE therefore represents how concentrated the data is around the line of best fit. Applied to this paper, the line of best fit would correspond to a perfect match between porosity measured with the CT-scan and that measured by AGAT laboratories using a conventional helium gas porosimetry. RMSE was used to give an idea of how well the CT-scan porosity matches the gas porosity obtained conventionally.

1) Helium porosity

The results allow for comparison of gas porosity obtained from three distinct instruments and laboratories (Tab. 2; 4; Fig. 8A-C). All three methods are simple and rapid techniques widely used to measure porosity on core samples. The porosity obtained from Boyle's law gas porosimeters (Fig. 8A and B), IHAP, AP-608 and AccuPyc, shows linear relationships with a R^2 coefficient ranging from 0.98 to 0.99 and a slope close to 1 (from 0.97 to 1.03). Accupyc and AP-608, with both measurement made at INRS, show the smallest RMSE (0.43%) when compared with IHAP (0.8% for Accupyc and 0.94% for AP-608). The larger differences between AP-608 and Accupyc (Fig. 8C) occur for limestone samples (Indiana and Carthage) that are known to be genetically more complex and spatially heterogeneous rocks (Galaup et al., 2012; Freire-Gormaly et al., 2015). Therefore, differences in porosity evaluation were expected. The same trend is observed when comparing IHAP with Accupyc and AP-608. This specific carbonate sample has a highly irregular surface (Fig. 9), with visible vugs and dissolved bioclasts on the exterior surface. Rock texture at surface can affect basic physical measurements such as length, particularly those made with a caliper, and in turn, can induce uncertainty in porosity estimation. Based on these comparison, results from sandstone samples appears to be more consistent. Analysis obtained from AGAT Laboratories were chosen as the most reliable since this private and independent laboratory runs routine petrophysical properties analyses (including helium gas porosimetry) for the private oil and gas sector as well as for academic and governmental research.

2) CT-scan porosity - reference samples

CT-scan porosity results are compared to IHAP only as the later values were validated using the two different instruments and can therefore be considered as a reference (Fig. 10). The porosity obtained by the IHAP and CT-scan method ($n=30$) shows a linear relationships with a R^2 coefficient of 0.99 and a slope close to 1 (0.91). When comparing CT-scan-IHAP with AP-608-IHAP, RMSE is more than two times higher (Fig. 8A). This difference is significant and tends to increase as porosity increases (Fig. 10). That can be explained by the fact that more porosity means more water present inside the sample thus a stronger beam hardening artefact variation. It leads to a larger underestimation of the porosity as beam hardening creates a greater underestimation of the density in the saturated state data.

A correction factor ($1/0.91$) was calculated using linear regression to minimize that difference (Fig. 10). This correction brings the RMSE down to 0.54% (Fig. 11), which is lower than AP-608 versus IHAP RMSE value (Fig. 8A), but slightly higher than AccuPyc versus AP-608 RMSE (Fig. 8C). This correction factor was applied to all data points and for all subsequent analyses. As already experienced by Ketcham and Iturrino (2005), some voxels have estimated porosity values below 0 % (and above 100 %), due to image noise and possible remaining misfit between data sets.

When considering each lithology type (Fig. 12), the CT-scan method seems to be less robust with dolomite rock samples, where R^2 decreases down to 0.83 (Fig. 12C) and RMSE is slightly higher. More data points would be needed to fully analyze these relationships. The influence of porosity range has then been considered (Fig. 13) and porosities values obtained using medical CT appears to be equally valid for the entire

range of porosities tested (from 1.5 to 34%), with R^2 around 0.98 (Fig. 13) and RMSE close to 0.5. The possible influence of core diameter was also investigated.

When comparing porosity results of 1^{1/2}" and 1^{7/8}" core diameter (Fig. 14), R^2 are very similar, with 0.99 (Fig. 14A) and 0.98 (Fig. 14B) respectively. However, RMSE increases from 0.43 to 0.72 %, from small to large diameter which seems to indicate that, even after linear correction, CT-scan method seems to produce more reliable porosity estimation for smaller diameter samples.

Lastly, CT-scan porosity results using doping agent are also compared with IHAP (Fig. 15) and show a linear relationship with a R^2 coefficient of 0.99 and RMSE equals to 0.70%. These results are comparable to those obtained with water saturation and the benefits of using NaI seems less important than initially expected.

3) CT-scan porosity – Silurian core interval

The selected HTD core interval (Fig. 4A) was subsampled and a 1.8 m continuous section was scanned using the core-flooding setup. The interval actually corresponds to 18 core subsamples (see Fig. 4 and Fig. 16). Within this 1.8 m interval, five isolated samples were also sent to AGAT laboratories to validate locally the CT-scan porosity values (Fig. 4B; namely CSI-2, 3, 10, 12 and 16). Small gaps between core sub-samples were taken into consideration and depths were corrected to account for these gaps. Different statistical profiles describing porosity were generated (Fig. 16) with a spatial resolution of 0.6 mm. Mean porosity profile and heterogeneity profile indicate that porosity is lower in the upper part of the depth interval but more heterogeneous, with a rather sharp transition between subsamples CSI-7 and CSI-8 (Fig. 16). Another interesting observation is the

occurrence of increased porosity intervals (with values higher than 10%) that seems to be limited to 10-15 cm thick interval (see for example within CSI-12 or 18). Such information would be completely missed if considering discrete samples only for helium gas porosity measurements. In addition to provide an average porosity value, CT-scan porosity dataset also provides valuable spatial information about the porosity within samples and allows porosity visualization in 3D (Fig. 17). Looking at diverse 3D views from each sample individually, qualitative information is added, such as porosity distribution vertically and horizontally, at the centimetric scale. For instance, within sample CSI-2, largest macropores are homogeneously distributed within the sample and correspond to isolated vugs (Fig. 17A), whereas in sample CSI-5, largest macropores are limited to specific areas of the sample, associated to oblique fractures (Fig. 17B).

DISCUSSION

1) Porosity interpretation

Reference samples

Overall porosity correlation is a bit better for sandstone samples (Fig. 12). However, RMSE for all lithologies remain quite close (and all $<0.7\%$). RMSE increases slightly from SST to LST and then Dol, a trend that coincide with porosity data points grouped more tightly (around 10% porosity approximately). The number of samples changes from $N=14$ to $N=6$, which can further affect the RMSE estimation. Even though our study use a large number of samples ($n=32$) compared to the available literature, it is obvious that the number of samples influences the quality of the results in a statistical point of view. In future works, we plan to extend porosity range for limestone and dolomite to have truly comparable dataset. However, the better porosity correlation for sandstone could very be

statistically valid and genetic in nature, and this well-known in the literature (Lucia, 2007; Bust et al., 2011; Victor et al., 2017). Carbonates and dolomites, because of their chemical reactivity have more complex diagenetic history and hence porosity distribution. Because of their physical properties and generally lower chemical reactivity, clastic sediments have more homogeneous porosity distribution to the contrary of carbonates in which heterogeneous distribution of calcitic, aragonitic and dolomitic components (particle, cement) will lead to irregular distribution of reactive particles to a specific fluid, and hence variable, even erratic, porosity development at the very fine scale.

Silurian core interval

The porosity evaluation of the Silurian core interval using CT-Scan (Fig. 16) gives clearly heterogeneous values as expected given the nature of samples. The core encompasses a mix of lithologies that do not appear affected by fracture-controlled hydrothermal fluid circulation (Fig. 3A) and intervals with diverse degrees of hydrothermal alteration (see Fig. 3B). The whole core set is a mix of preserved depositional limestone facies (non-porous low energy depositional environment represented by lime mudstone and wave reworked porous bioclastic limestone) and hydrothermally altered diagenetic facies.

The improved methodology using medical CT-scan allows mapping out the porosity at a centimetric scale, something impossible to achieve through conventional approaches. The added value of the improved methodology is well illustrated when comparing dry state CT-scan images of two dolomitic samples with similar porosity value (around 9.5 %; Fig.

18; Table 4-5). A reference sample (SI-K-15A) presents circular to ovoid mesopores mainly concentrated on the lower half of the specimen (Fig. 18A). Macro/mesopores appear disconnected from one another, at least at the medical-CT scale resolution. The core sample coming from the silurian hydrothermally altered interval is completely different (Fig. 18B), even though core-averaged porosity value are very close. Macropores are much larger and their distribution is highly heterogeneous. Such qualitative information cannot be deduced from conventional helium porosity. In addition, when these two samples are analysed by our improved methodology, core-averaged porosity values are different. The reference samples, Si-K-15A, has a lower porosity estimates (8.31%) whereas the silurian sample hydrothermally altered has a higher porosity estimates (11.3%). Looking at the images (Fig. 18), the CT-scan, core-averaged, porosity values appears to be more realistic.

2) Advantages of this new CT scan methodology

The average CT porosity value for each reference sample has been used to validate the methodology but the ultimate goal is to access the spatial distribution of porosity and acquire further information on porosity at the macro scale. To achieve that, 2D and 3D visualisation can be performed on the dataset to qualitatively describe the porosity distribution. Quantitative statistics can also be derived to better describe not only porosity values but also its spatial distribution: pore concentration at specific levels, heterogeneity variation with function of depth, etc.

The experimental setup developed in the present study is cost-effective and easy to handle (Fig. 5), especially when compare to previous core-flooding experiments found in the literature (Hove et al., 1987; Vinegar and Wellington, 1987; Withjack, 1988), often using

costly pressure vessels made of aluminium chamber and Teflon casing (i.e. Hassler type core holder; (Karacan et al., 2003). Different from Ketcham and Iturrino (2005) , our current setup presents no risk of fluid loss while saturation and scanning phases because samples stay within the core holder at all times. Numerical realignment of dry and wet states allows to perform samples saturation phase (a minimum of 72 hours duration) outside of the CT scanner requiring the use of the CT facility for a very limited amount of time. The low costs of the core holder (PVC or Plexiglas) and custom, 3D printed internal fittings (PLA), makes it easy to design a decimetric core-holder and analyse several samples at the same time. As a matter of fact, the cost-effectiveness of both setup and protocol allowed to perform CT-scan based porosities determination on 30 isolated samples and a 1.6 m thick core section. This is, to the best of our knowledge, the first attempt ever made to run core-flooding experiments under CT on such a large number samples (Tab. 1; Fig. 1). The scan time is quite reasonable (2 minutes for 3 repeated scan on a 10cm sample), data processing time is fast (6 minutes) and the method is scalable to process tens of meters of core samples. The saturation process was performed for a week as the maximum water saturation was desired but this process could be optimized by adjusting the following variables: (1) degassing time, (2) vacuum and water circulation time and (3) water circulation duration without vacuum (Fig. 6). The second step takes most of the process time and might be reduced considering the porosity level of the sample. It might also be desirable to spend more time on the step 3 as remaining air bubbles shrink in size and thus allow more water in. Likewise, it might be more effective to perform alternative vacuum during step 2, as this could split entrapped air into smaller bubbles which could then be extracted from the sample.

Different approaches used to derive quantitative information from CT data needs calibration. Converting HU to density profiles requires subsampling, volume and weight measurements

(Boespflug et al., 1994; Amos et al., 1996; ASTM-E1441-11, 2011). Mineralogy and porosity variation obtained from “CT dual energy” acquisition are results derived from effective atomic number and density profiles, which in turn needs calibration from materials standards to convert X-ray absorption measurement (Van Geet and Swennen, 2001; Walls and Armbruster, 2012; Lopez et al., 2016). Calibration is a non-trivial process and cumulates errors from each additional required measurement. The proposed approach includes calibration, as it requires air and fluid HU values, but is straightforward and taken in situ, for every measurement made. Air HU value is measured while scanning samples at dry state and fluid HU value during the saturated state scans.

No prior knowledge about mineralogy, density or porosity range is needed with the present methodology to obtain a reliable porosity value (Figs. 10-16; Table 4-5). CT-scan porosities obtained for dolomite samples are little less correlated with porosities values obtained by conventional porosimetry, but overall lithology type still has a low influence on results (Fig. 12).

One of the main difference between porosities derived from CT-scan versus those given by gas porosimeter is its independence from volume calculation. Gas porosimeters measure grain volume. The porosity is then calculated using bulk volume of sample which is based on linear measurements of samples with a caliper and the application of the appropriate geometric formula. Therefore, this method is subject to human error and measurement error if the sample is irregularly shaped (e.g. Fig. 9).

Lastly, working at medical CT scale represents an advantage when considering scanning time and sample size. With our working parameters (Tab. 3), acquisition time is few seconds to minutes, compare to micro or nano-CT where total data acquisition times span from hours to

ten of hours (Cnudde and Boone, 2013; Bultreys et al., 2016a) and where sample size is often limited to only few millimeters (Pini and Madonna, 2016), which brings back the question of the representativeness.

3) Limitations

One limitation of the proposed approach is the need to further correct beam hardening effects. The original postulate was that artefacts such as beam hardening do not need to be corrected because (1) the geometry is cylindrical (Pini and Madonna, 2016) and (2) remains constant through flooding experiment, i.e. beam hardening effect would be canceled out via the subtraction of wet and dry datasets. However our results have shown that the difference increases linearly as the porosity increases (Fig. 8). More porous samples (therefore associated with the greatest difference between the two states), are associated with lower correlations with gas-measured porosity values. Nonetheless, our current dataset suggests that a simple linear correction seems to account well for the residual influence of beam hardening (Fig. 10-11).

Cylindrical samples ease the selection of a versatile and low cost core holder material that would then accommodate a large range of sample diameter. However the restrictive geometry itself could be seen as a limiting factor. Without proper and full beam hardening correction, the analysis of randomly shape rock fragments and other sample geometry remains difficult under CT-scan. At best, qualitative information could be obtained from non-cylindrical specimen.

4) Perspectives on future work (s)

Future research efforts should explore whether the experimental protocol, the acquisition parameters or the data analysis itself can be optimized by lithology type or porosity

range. Maybe one or two repeated scans would be enough and therefore could lower the cost to some instances.

Another key point for further works is to test upscaling possibilities. In particular, the calibration of wireline logs profiles using the CT-scan porosity profiles, instead of discrete porosity values from plugs, is promising. CT-scan images can be correlated directly with density well log because they both measure the amount of Compton scattering, proportional to bulk density (Wellington and Vinegar, 1987). With the high levels of heterogeneity inherent in carbonate reservoirs, correlation between low resolution e-logs and high resolution, discrete, poro-perm measurements has been debated (Delhomme et al., 1996; Tilke et al., 2006). The medical CT and its range of investigation could well bridges the gap.

As already stated, the production of 3D porosity matrix images (e.g. Fig. 17) opens the opportunity to produce 3D models and run numerical flow simulations, at the centimetric or tens of centimeter scale. This would be of interest for many research fields such as oil, gas and geothermal reservoirs, hydrogeology or CO₂ sequestration.

CT-scan images were only made at the initial (samples filled with air) and final stages (samples filled with distilled water), therefore no information regarding transitional saturation conditions, wetting characteristics of rocks or permeability were gained. Compared to similar studies in this field of research (Vinegar and Wellington, 1987; Wellington and Vinegar, 1987; Withjack, 1988), there is no gas/fluid front to track. Furthermore, the current setup rely only on capillary forces to saturate the rock samples. Said differently, no fluid forcing is applied. One can argue that porosity data alone are

insufficient. Two samples with similar porosity values can have significantly different permeability. However, the current setup not only allows the assessment of an average porosity value per sample, but also provides a 3D porosity matrix (Fig. 17). This in turn, can be transformed into 3D models and used to run numerical flow simulations using commercial software such as COMSOL. Such approach has already been tested with micro-CT and medical CT measurements (e.g. Zaretskiy et al., 2010; Bultreys et al., 2015). To our knowledge, this has not commonly been achieved on heterogeneous carbonate samples, for which pore structure measurements are mostly based on mercury injection (Galaup et al., 2012), eventually combined with micro-CT for low porosity carbonate (Fusi and Martinez-Martinez, 2013). A lot of work has been done in terms of fluid flow modelling and simulation (see Review papers by Meakin and Tartakovsky (2009) ; Blunt et al. (2013) and references therein) and might be adapted to our (medical) CT-scan porosity dataset.

CONCLUSION

This work developed an effective and practical method using medical-CT to reliably estimate reservoir porosity for spatially heterogeneous material such as fractured or dolomitized carbonates, incorporating recent advances in data correction.

(1) The in-house core-flooding setup is low cost, simple, and easy to operate. Several individual core samples can be scanned simultaneously (dry and saturated), as well as continuous core sections up to 1.5 m long. Scanning a sample in a dry state and saturated state, performing a three-dimensional alignment and subtracting the two data sets allow the construction of 3D porosity matrices.

(2) Based on a set of reference core material, this study illustrates the relationship between porosity assessed by CT-scan against the ones obtained by conventional gas porosimetry techniques. A strong correlation is observed between both techniques so that the current CT-scan methodology appears to be a reliable and accurate way to estimate fine-scale variations of porosity for the main types of sedimentary rocks, in a wide range of porosity value.

(3) This consistency opens up the possibility to extend porosity assessment beyond gas porosimetry, particularly for heterogeneous carbonate samples. The added value of the porosity measurement by CT-scan is the generation of 3D images of pore network, allowing to assess spatial attributes of macropores, their distribution and connectivity.

(4) Last, but not least, the CT-scan method allows the construction of continuous porosity profiles that are well correlated with discrete helium gas porosity values. Millimetric/centimetric scale data are rarely available in subsurface datasets and reliable continuous porosity measurement at this scale is a step forward in the understanding of reservoir properties.

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FIGURES CAPTIONS

TAB. 1 List of references samples used for the present work. In the sample name, the petroleum core sample provider is indicated, with K standing for Kocurek Industries Inc and C standing for Cleveland Quarries. All basic samples measurements were performed at the LOG, using a digital caliper and a precision scale. BE stands for Berea sandstone; SC stands for Scioto sandstone; BO stands for Boise sandstone, NU stands for Nugget sandstone; IN stands for Indiana sandstone; CA stands for Carthage Marble (= Burlington Limestone); GE stands for Guelph dolomite; SI stands for Silurian dolomite.

TAB. 2 Comparison of methods used in the present work to estimate porosity.

TAB. 3 Summary of CT-scanner parameters values for both acquisition and reconstitution stages. kV stands for kilovoltage, mAs for milliampere-second, F.O.V. for field of view, and HU for Hounsfield Unit.

TAB. 4 Summary of results for reference core samples. The absolute error (AE) corresponds to the difference between the porosity calculated using the improved CT-scan methodology and the porosity measured using the conventional gas porosimeter IHAP.

TAB. 5 Summary of results for silurian core samples.

FIG. 1 Range of porosity tested for this work. Reference core samples correspond to eight different lithologies commonly used as test material in the petroleum industry, and cover a large range of porosity (2-5 to 28 %), namely Berea, Scioto, Nugget and Boise sandstones (SST), Indiana and Burlington limestones (LST), and Silurian and Guelph dolomites. For each lithology type, an expected porosity or porosity range was given by the vendor Kocurek Industries and these values are reported here.

FIG. 2 Petrographic attributes of reference core material. (A) Microphotograph of medium-grained Nugget sandstone sample, with moderate sorting and well-rounded grains. (B) Microphotograph of medium to coarse-grained Boise sandstone sample, with poor sorting. Quartz grains are angular (C-D) Microphotographs of fine-grained, well sorted Scioto sandstone. Quartz grains are subangular. (E) Microphotograph of Carthage marble limestone. This limestone is a well cemented, fossiliferous limestone with moderate to poor sorting. (F) Microphotograph of homogeneous fine grained Silurian dolostone within the Sayabec Formation.

FIG. 3. Simplified geological map of the Témiscouata area in eastern Quebec (Canada) with the location of the Massé No 1 well (black star). The yellow star locates the position of an exhumed hydrocarbon field hosted in hydrothermal dolomite (HTD). The stratigraphic column to the left locates the Sayabec - Val Brillant interval deposited at the end of the first shallowing event (S1) and onset of the first deepening event (D1) in the Gaspé Belt. Stratigraphic details and basin evolution are found in Bourque et al. (1995).

FIG. 4. (A) Macrophotograph of silurian core interval used for this study. These continuous 4.5 meter long core section belong to the lower Silurian Sayabec Formation in Massé No. 1 well drilled within the Massé structure (Lower St-Lawrence river area, Québec). Core samples are 4.5 cm in diameter and their length ranges from 5 to 10 cm (approximately). (B) Macrophotograph of five Silurian core samples, chosen to illustrate the natural heterogeneity of this interval along depth. Each sample is briefly described and porosity is assessed from macroscopic observations on the surface.

FIG. 5 Schematic diagram of the core flooding experimental setup for porosity measurement. The water tank is a closed reservoir with a 4 L total volume. To accommodate meter long core section, four chamber are set in parallel.

FIG. 6 Workflow chart illustrating the successive steps involved in the present work and separated in three groups: experiment, CT-scanning acquisition and processing.

FIG. 7 (A) Impact of noise level on porosity calculation and its uncertainty level (adopted from Pini and Madonna (2016)). (B) Axial CT-scans with decreasing resolution. This illustrates how fine structures (such as fractures) could remain undetected if the resolution is too low. The spatial resolution was then set to 0.1 x 0.1 x 0.6 mm.

FIG. 8 Statistical comparison of three gas porosity measurement techniques. (A) IHAP versus AP-608 with $n = 20$, (B) IHAP versus AccuPyc with $n = 10$, and (C) AP-608 versus AccuPyc, with $n = 10$.

FIG. 9 Indiana limestone sample (IN-C-178B) with its highly irregular surface. Macropores and core damages are abundant on the external surface which produce an imprecise total volume calculation using caliper and could induce the outlier data point (see Fig. 6; Fig. 9A).

FIG. 10 Statistical comparison of CT-scan porosity measurement technique against conventional gas porosity technique (IHAP). The outlier results (white star) corresponds to carbonate sample IN-C-178B, and was not considered for regression. For further analyses and subsequent figures, the slope of the regression line is used as a correction factor.

FIG. 11 Statistical comparison of CT-scan porosities against IHAP after correction. The correction factor used (1/0.91) intends to correct beam hardening effect. The outlier carbonate sample (IN-C-178B) and was not considered for regression.

FIG. 12 Lithology influence on the correlation between CT-scan porosity method and conventional IHAP. Note that all data point used are corrected values. (A) Data points for sandstones (n=14); (B) data points for limestones (n=10) and (C) data points for dolomites (n=6). Indiana limestone outlier (IN-C-178B) was not considered for regression.

FIG. 13 Porosity range influence on the correlation between CT-scan porosities and conventional IHAP. Note that all data point used are corrected values. Low porosity range is defined as porosity values lower than 15 % (n=10); and high porosity range is defined as porosity values higher than 15% (n=19). Indiana outlier (IN-C-178B) was not taken in consideration for regression. R^2 for both ranges of porosity values reaches 0.98, and RMSE are rather low, close to 0.5.

FIG. 14 Core diameter influence on the correlation between CT-scan porosities and conventional IHAP, with (A) 1½" diameter core samples and (B) 1^{7/8}" diameter core samples.

FIG. 15 Statistical comparison of CT-scan porosities obtained with doping agent (NaI) against IHAP porosities. Note that all data point used are corrected values. Two outlier samples (not shown) were not considered for regression. Both outlier correspond to Indiana limestone samples, and one of them is IN-C-178B (Fig. 8).

FIG. 16 Continuous porosity profiles obtained for a 1.8 meters thick section of the lower Silurian hydrothermal dolomites. The average porosity for this entire section is 3.38%.

The continuous porosity values are compared to discrete helium gas porosity values obtained at the AGAT laboratories, and CT-scan porosity value average per subsamples (CSI-3, CSI-6, CSI-10, CSI-12 and CSI-16). The different profiles were plotted using IP software. A bell (Gaussian) filter was applied to the mean CT-scan porosity values. HI (heterogeneity index) and CV (coefficient of variation) parameters are adopted from Caliskan and Shebatalhamd (2017)

FIG. 17 Examples of 2D views illustrating connected porosity matrices obtained for two specific subsamples within the HTD interval. Greyscale indicates porosity, from 0 % (black) to 100% (white). (A-B) correspond to coronal and sagittal mean intensity projections views (MIP) respectively of the sample CSI-2 Sample CSI-5 MIP views are shown in the same manner in (C-D). See Fig. 15 for samples location. Provided as supplementary material, 360° rotation movies of these two samples were made using Dragonfly software.

FIG. 18 Examples of coronal CT images from (A) a reference core sample, namely SI-K-15A and from (B) one sample from the HTD interval, namely CSI-12 (see Fig. 16 for its location along the porosity profile). Both dolomite samples have similar porosity values obtained by IHAP, i.e. 9.5% and 9.32 % respectively (see Table 4 and 5).

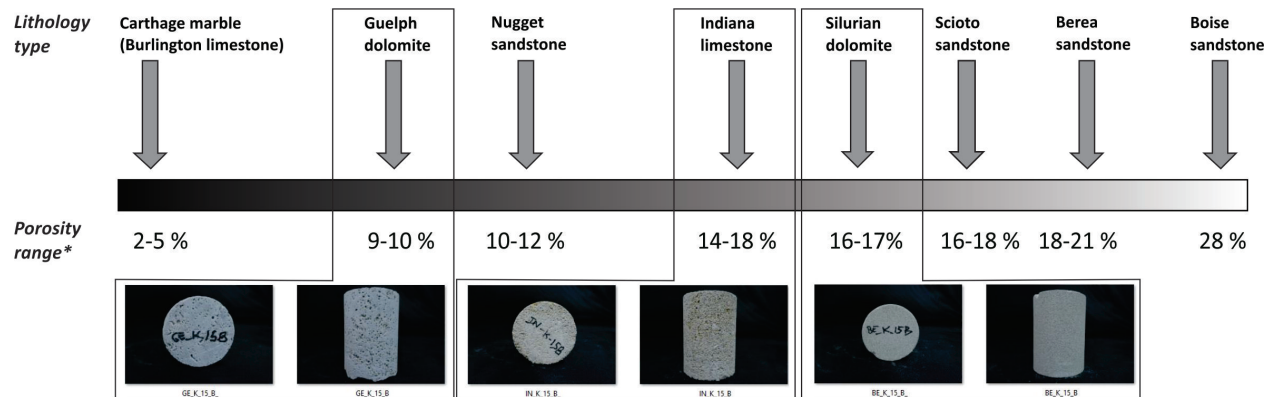
| Location | Number of specimen | Lithology type | Average porosity (from literature) | References |
|---------------------------------|--------------------|-------------------|------------------------------------|--|
| Ohio (USA) | 8 | Berea SST | 20% | Winkler, 1983; Churcher et al., 1991; Hart and Wang, 1995; Boon et al., 2017 |
| Utah and Wyoming (USA) | 2 | Nugget SST | up to 25% | Lindquist, 1988 |
| Idaho (USA) | 2 | Boise SST | 28-30% | Winkler, 1983 |
| Ohio (USA) | 2 | Scioto SST | 12% | Holder et al., 2001; Bose et al., 2014 |
| Indiana (USA) | 8 | Indiana LST | 13% | Musselman, 1967; Schmidt and Huddle, 1977; Churcher et al., 1991; Hart and Wang, 1995; Boon et al., 2017 |
| Ohio (USA) | 2 | Carthage LST | 1.5% | Musselman, 1967; Martin, 1968 |
| Ohio (USA) | 2 | Silurian dolomite | 14-20% | Islam et al., 2018 |
| Ohio (USA) and Ontario (Canada) | 4 | Guelph dolomite | 7-24% | Churcher et al., 1991 |

| | AccuPyc | AP-608 porosimeter | AGAT Helium porosimeter | CT-scan |
|--------------------------|---------|--------------------|----------------------------|-----------------------|
| Sample length (cm) | 2.54 | 2.54 to 10.16 | 2.54 to 7.62 | up to 250 |
| Sample diameter (cm) | 2.54 | 2.54 | 2.54 or 3.81 or 5.08 | up to 50 |
| Injection pressure (psi) | 20 | 200 | 100 | <i>not applicable</i> |

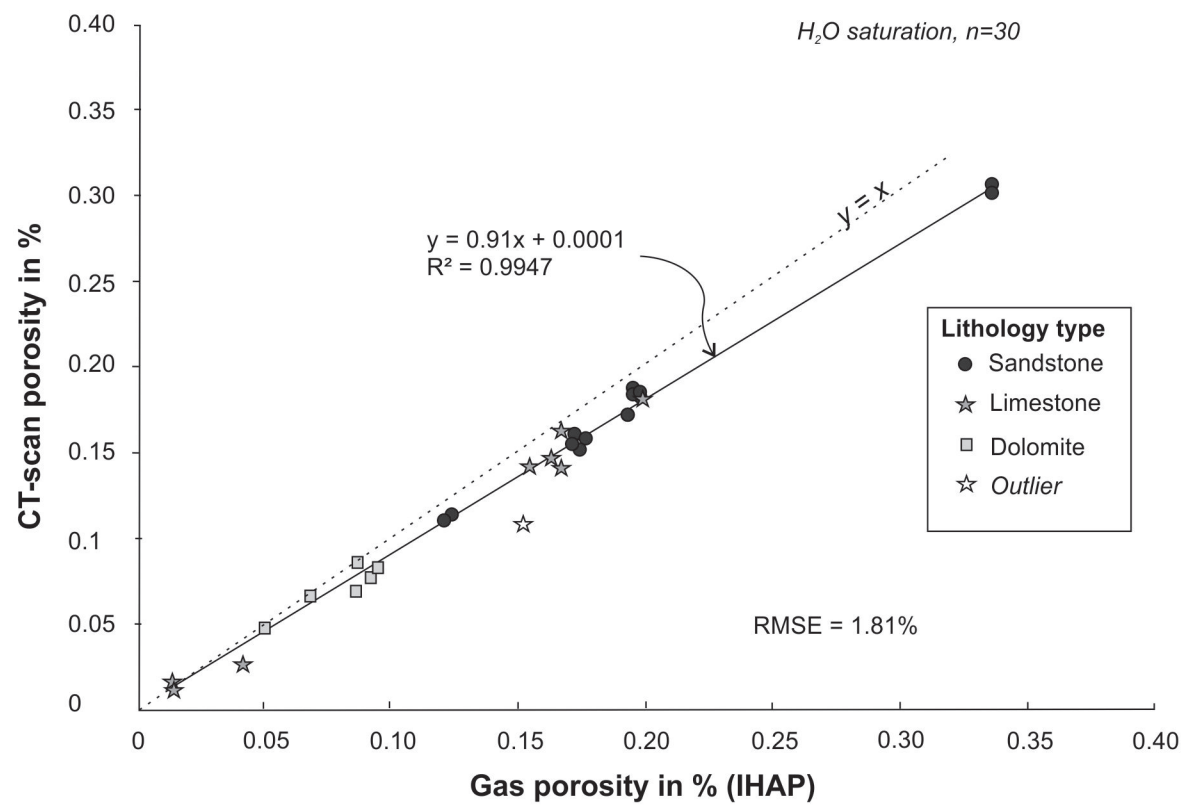
| Acquisition parameters | | |
|---------------------------|-------------------|----------------------|
| | Reference samples | Natural core samples |
| kVp | 140 | 140 |
| mAs | 700 | 350 |
| Pitch | 0.55 | 0.55 |
| Collimation | 20 x 0,6 mm | 20 x 0,6 mm |
| Reconstruction parameters | | |
| | Reference samples | Natural core samples |
| Filter | H70h | H70h |
| F.O.V | 60 mm | 55 mm |
| Pixels spacing | 0.1172 x 0.1172 | 0.1074 x 0.1074 |
| Slice thickness | 0.6 mm | 0.6 mm |
| HU scale | normal | normal |
| Focal spot | 1.2 mm | 1.2 mm |

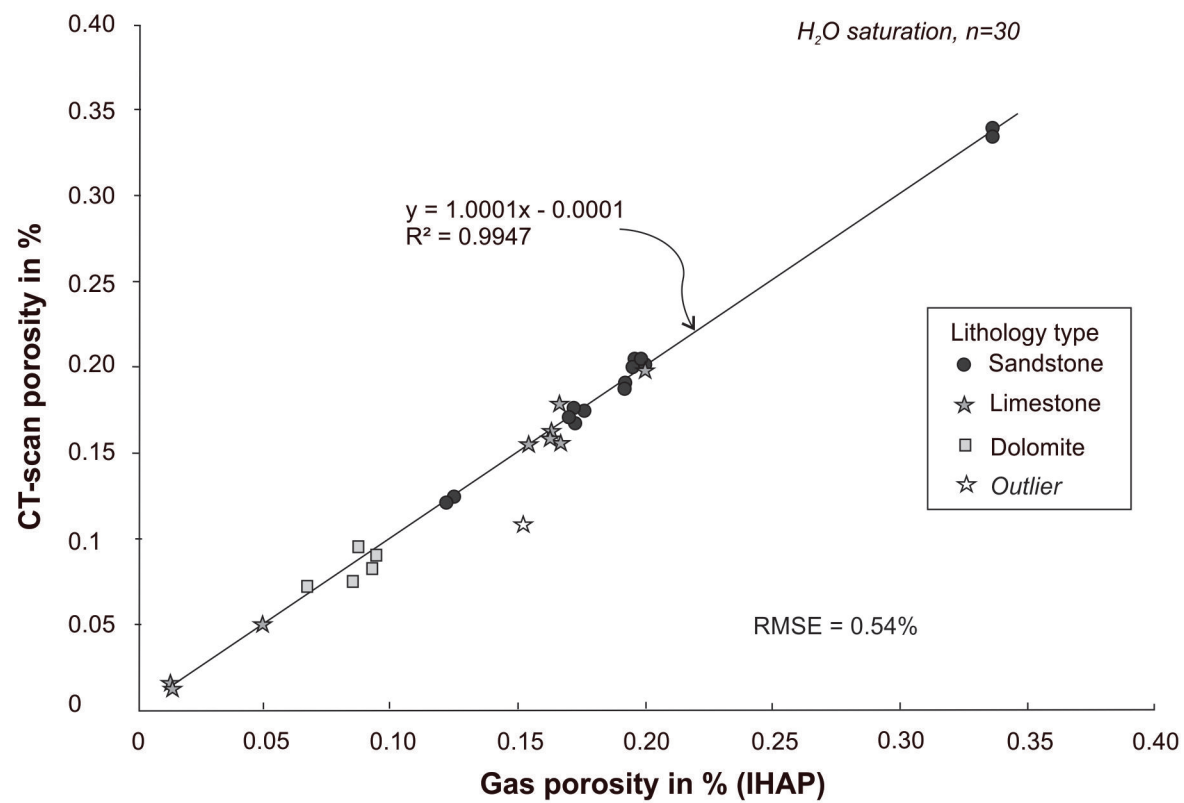
| Sample attributes | | | | | Measured porosity (this work) | | | | | Absolute error | Reference porosity | |
|-------------------|-------------|---------------|------------|--------------|-------------------------------|--------|--------|---------------|---------------|----------------|--------------------|-------------|
| Sample name | Length (mm) | Diameter (mm) | Weight (g) | Lithology | AccuPyc | AP-608 | IHAP | CT-scan (H2O) | CT-scan (Nal) | | Littérature | Core vendor |
| BE_C_15_A | 9.9 | 3.8 | 236.00 | Berea SST | n/a | 20.30% | 19.80% | 20.02% | n/a | 0.22% | 20% | 18-21% |
| BE_C_15_B | 5.2 | 3.8 | 121.70 | Berea SST | 20.16% | 20.31% | 19.80% | 20.35% | 17.21% | 0.55% | | |
| BE_C_178_A | 74.9 | 4.5 | 287.87 | Berea SST | n/a | n/a | 17.10% | 17.12% | n/a | 0.02% | | |
| BE_C_178_B | 75.2 | 4.5 | 290.12 | Berea SST | n/a | n/a | 17.20% | 17.71% | 12.75% | 0.51% | | |
| BE_K_15_A | 9.9 | 3.8 | 238.30 | Berea SST | n/a | 19.74% | 19.30% | 18.95% | n/a | 0.35% | | |
| BE_K_15_B | 5.1 | 3.8 | 124.00 | Berea SST | 19.38% | 20.25% | 19.30% | 18.86% | 16.52% | 0.44% | | |
| BE_K_178_A | 74.6 | 4.5 | 311.50 | Berea SST | n/a | n/a | 19.50% | 20.38% | n/a | 0.88% | | |
| BE_K_178_B | 75.5 | 4.5 | 315.28 | Berea SST | n/a | n/a | 19.50% | 20.20% | 16.14% | 0.70% | | |
| NU-K-15A | 8.21 | 3.83 | 217.00 | Nugget SST | n/a | 13.27% | 12.10% | 12.20% | n/a | 0.10% | up to 25% | 10-12% |
| NU-K-15B | 4.97 | 3.84 | 130.80 | Nugget SST | 13.63% | 14.23% | 12.30% | 12.34% | 10.57% | 0.04% | | |
| BO-K-15A | 8.13 | 3.79 | 156.37 | Boise SST | n/a | 34.99% | 33.60% | 33.68% | n/a | 0.08% | 28-30% | 128% |
| BO-K-15B | 4.91 | 3.79 | 94.51 | Boise SST | 34.60% | 34.36% | 33.60% | 33.19% | 28.45% | 0.41% | | |
| SC-K-15A | 8.13 | 3.82 | 199.80 | Scioto SST | n/a | 19.45% | 17.40% | 16.75% | n/a | 0.65% | 12% | 16-18% |
| SC-K-15B | 4.93 | 3.81 | 121.20 | Scioto SST | 18.90% | 18.89% | 17.60% | 17.45% | 14.82% | 0.15% | | |
| IN_C_15_A | 9.90 | 3.77 | 248.50 | Indiana LST | n/a | 17.18% | 16.30% | 16.15% | n/a | 0.15% | 13% | 14-18% |
| IN_C_15_B | 5.10 | 3.77 | 127.70 | Indiana LST | 16.61% | 17.59% | 16.30% | 16.04% | 12.77% | 0.26% | | |
| IN_C_178_A | 74.36 | 4.50 | 294.95 | Indiana LST | n/a | n/a | 15.40% | 15.58% | n/a | 0.18% | | |
| IN_C_178_B | 76.31 | 4.50 | 304.64 | Indiana LST | n/a | n/a | 15.20% | 11.46% | 10.60% | 3.74% | | |
| IN_K_15_A | 9.80 | 3.76 | 233.30 | Indiana LST | n/a | 20.45% | 19.90% | 19.88% | n/a | 0.02% | | |
| IN_K_15_B | 5.20 | 3.76 | 123.20 | Indiana LST | 18.85% | 21.01% | 19.90% | 19.97% | 15.98% | 0.07% | | |
| IN_K_178_A | 74.38 | 4.50 | 325.99 | Indiana LST | n/a | n/a | 16.70% | 17.88% | n/a | 1.18% | | |
| IN_K_178_B | 75.50 | 4.50 | 331.30 | Indiana LST | n/a | n/a | 16.70% | 15.53% | 7.75% | 1.17% | | |
| CA-K-15A | 8.07 | 3.81 | 242.70 | Carthage LST | n/a | 2.76% | 1.40% | 1.37% | n/a | 0.03% | 1.50% | 2-5% |
| CA-K-15B | 5.01 | 3.818 | 150.4 | Carthage LST | 2.82% | 3.34% | 1.40% | 1.51% | 0.57% | 0.11% | | |
| SI-K-15A | 8.05 | 3.83 | 236.10 | Silurian Dol | n/a | 9.69% | 9.50% | 9.11% | n/a | 0.39% | 14-20% | 16-18% |
| SI-K-15B | 4.99 | 3.84 | 146.20 | Silurian Dol | 10.14% | 10.09% | 9.20% | 8.38% | 6.47% | 0.82% | | |
| GE_K_15_A | 8.95 | 3.78 | 260.00 | Guelph Dol | n/a | 8.96% | 8.60% | 7.65% | n/a | 0.95% | 7-24% | 9-10% |
| GE_K_15_B | 5.14 | 3.78 | 147.70 | Guelph Dol | 8.94% | 10.63% | 8.70% | 9.49% | 7.03% | 0.79% | | |
| GE_K_178_A | 66.91 | 4.50 | 355.17 | Guelph Dol | n/a | n/a | 5.00% | 5.22% | n/a | 0.22% | | |
| GE_K_178_B | 76.13 | 4.50 | 397.71 | Guelph Dol | n/a | n/a | 6.80% | 7.32% | 5.24% | 0.52% | | |

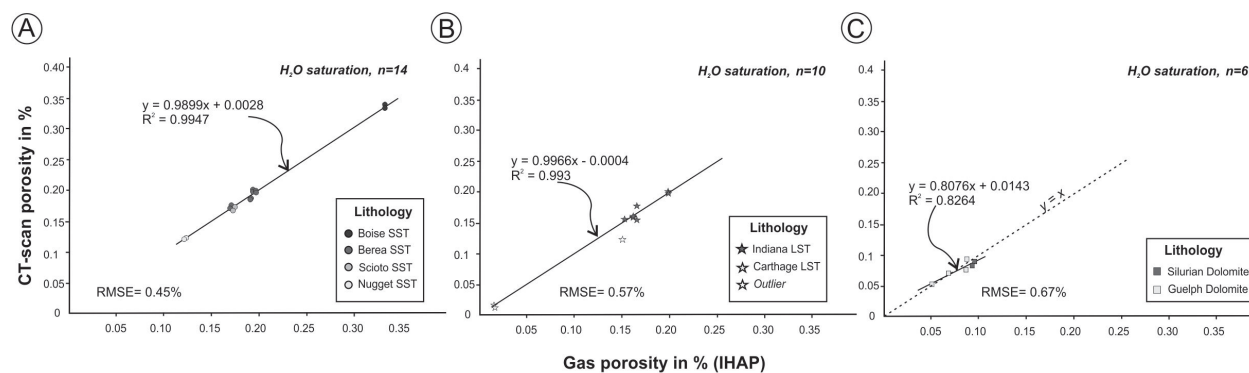
| Sample attributes | | | Measured porosity (this work) | |
|-------------------|-------------|---------------|--------------------------------|---------------|
| Sample name | Length (cm) | Diameter (mm) | CT-scan (H2O) <i>corrected</i> | IHAP |
| CSI-1 | 13.0 | 4.5 | 7.76% | n/a |
| CSI-2 | 9.0 | 4.5 | 7.03% | n/a |
| CSI-3 | 5.5 | 4.5 | 2.51% | 2.10% |
| CSI-4 | 12.5 | 4.5 | 1.93% | n/a |
| CSI-5 | 11.0 | 4.5 | 2.76% | n/a |
| CSI-6 | 6.0 | 4.5 | 2.42% | 1.90% |
| CSI-7 | 5.0 | 4.5 | 1.43% | n/a |
| CSI-8 | 18.0 | 4.5 | 6.82% | n/a |
| CSI-9 | 6.50 | 4.5 | 5.73% | n/a |
| CSI-10 | 7.00 | 4.5 | 4.54% | 4.20% |
| CSI-11 | 17.00 | 4.5 | 3.99% | n/a |
| CSI-12 | 11.00 | 4.5 | 9.36% | 11.30% |
| CSI-13 | 5.00 | 4.5 | 4.35% | n/a |
| CSI-14 | 11.00 | 4.5 | 8.89% | n/a |
| CSI-15 | 4.00 | 4.5 | 7.43% | n/a |
| CSI-16 | 6.00 | 4.5 | 2.53% | 3.00% |
| CSI-17 | 4.00 | 4.5 | 3.60% | n/a |
| CSI-18 | 9.50 | 4.5 | 2.92% | n/a |

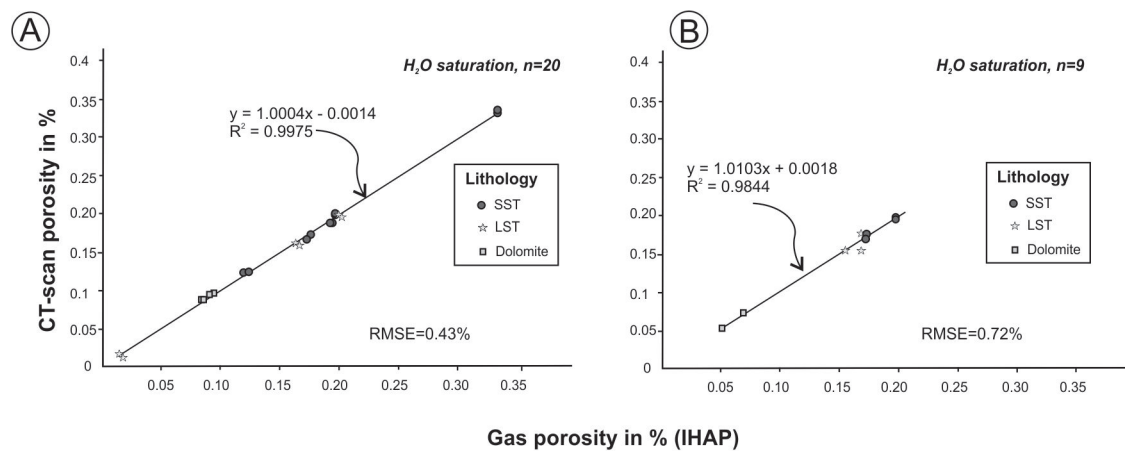


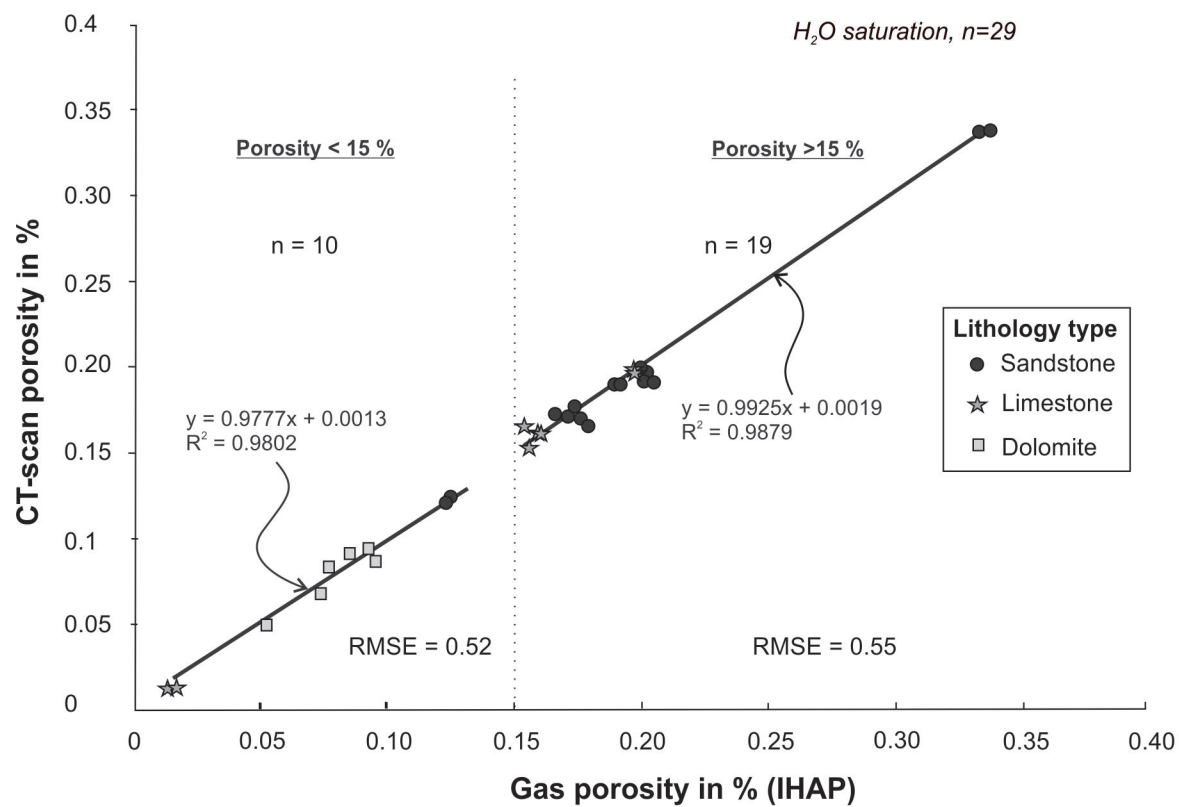
*source Kocurek Industries Hard Rock Division (<https://kocurekindustries.com>)

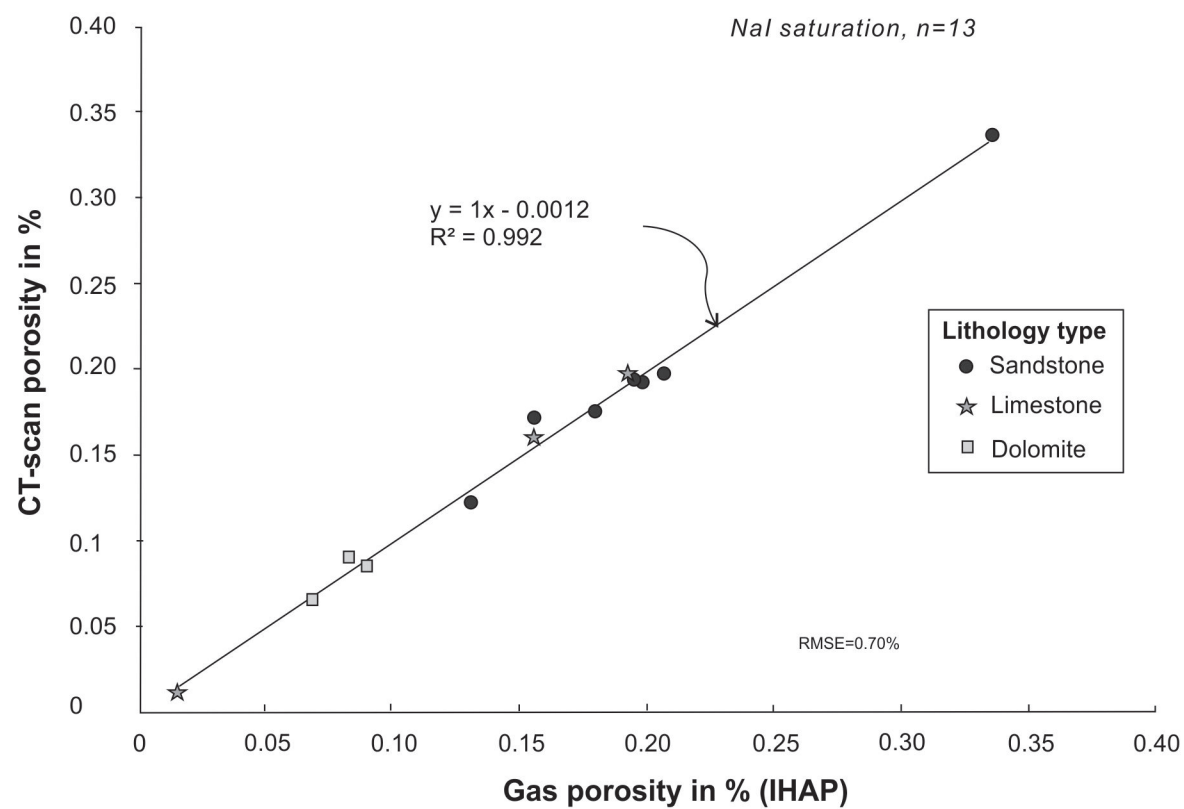


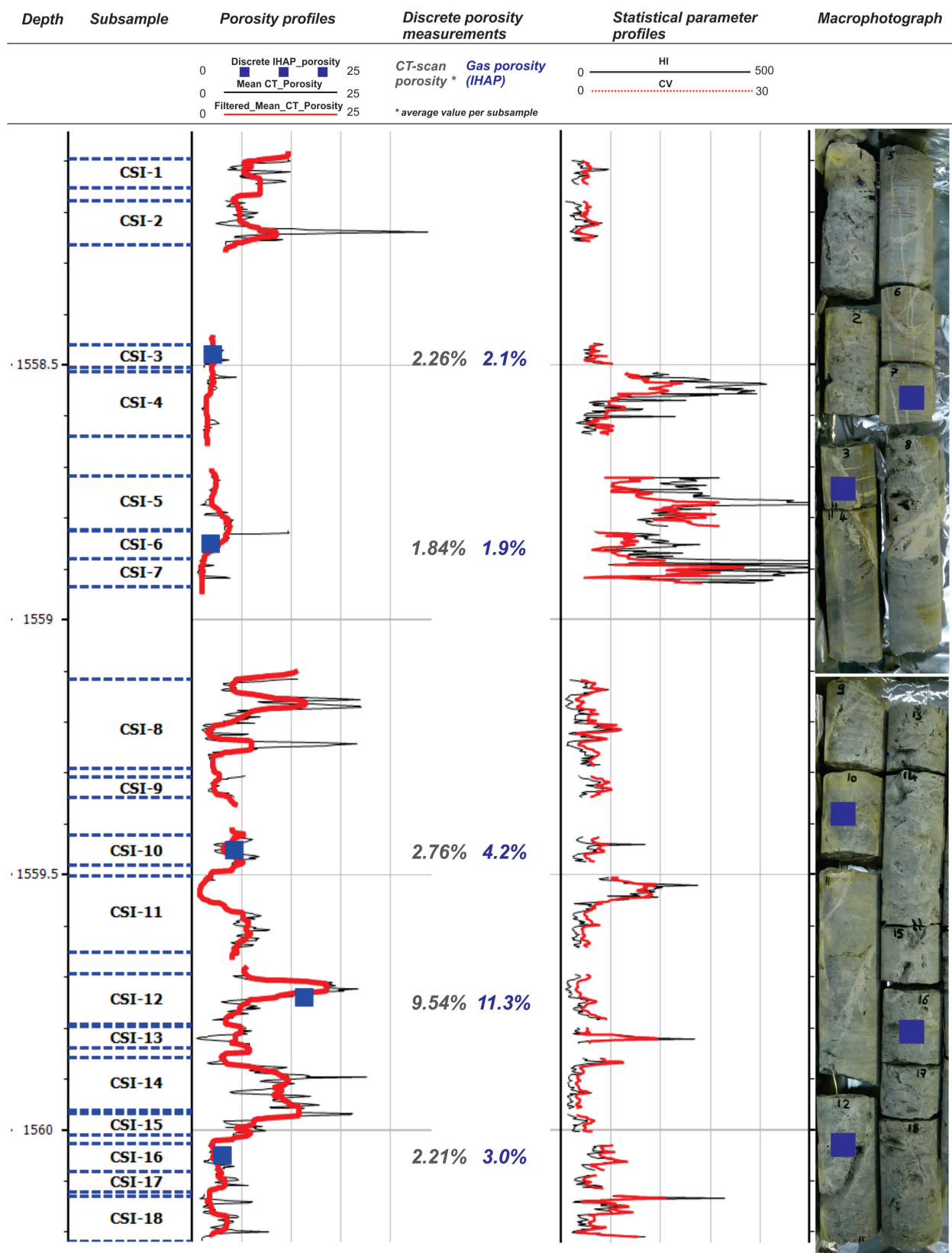


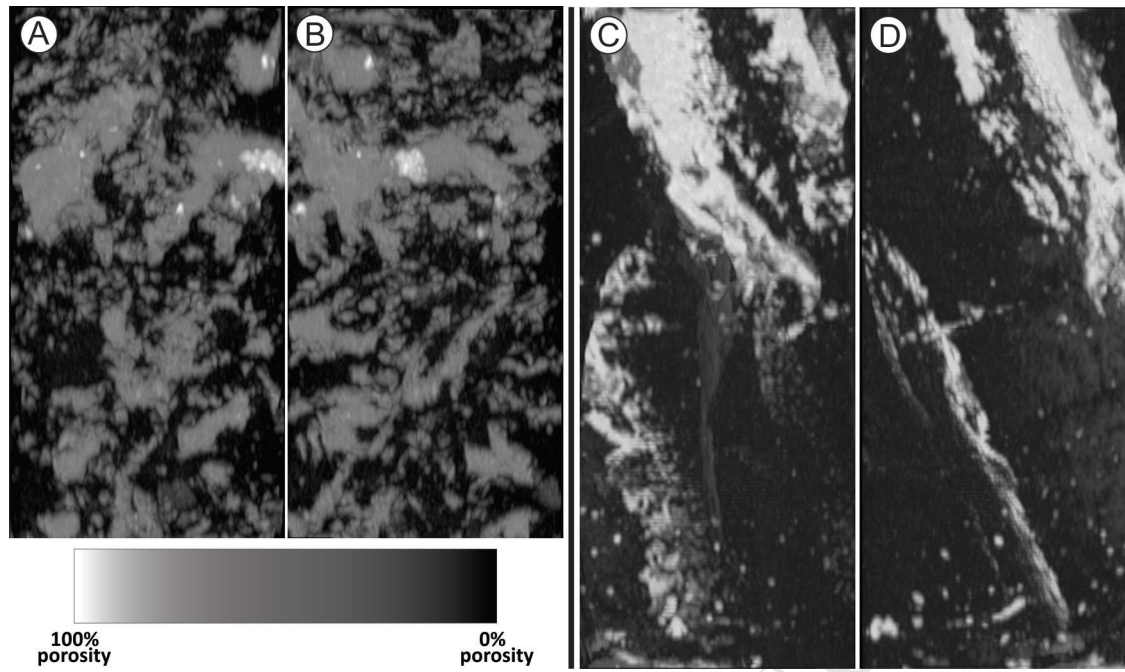


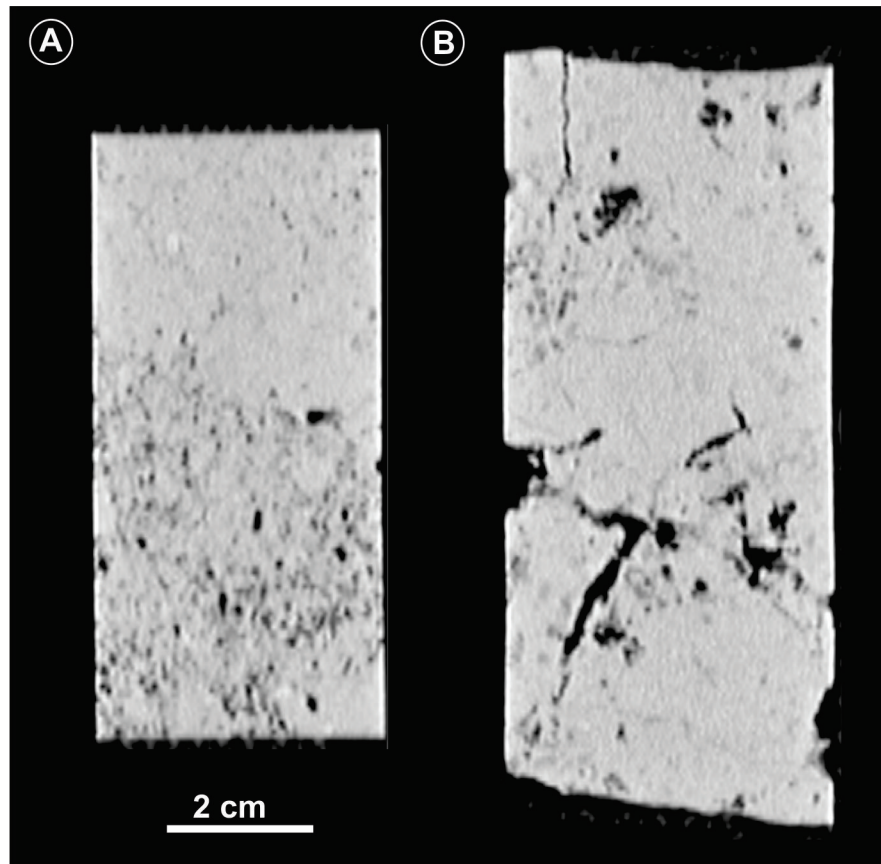


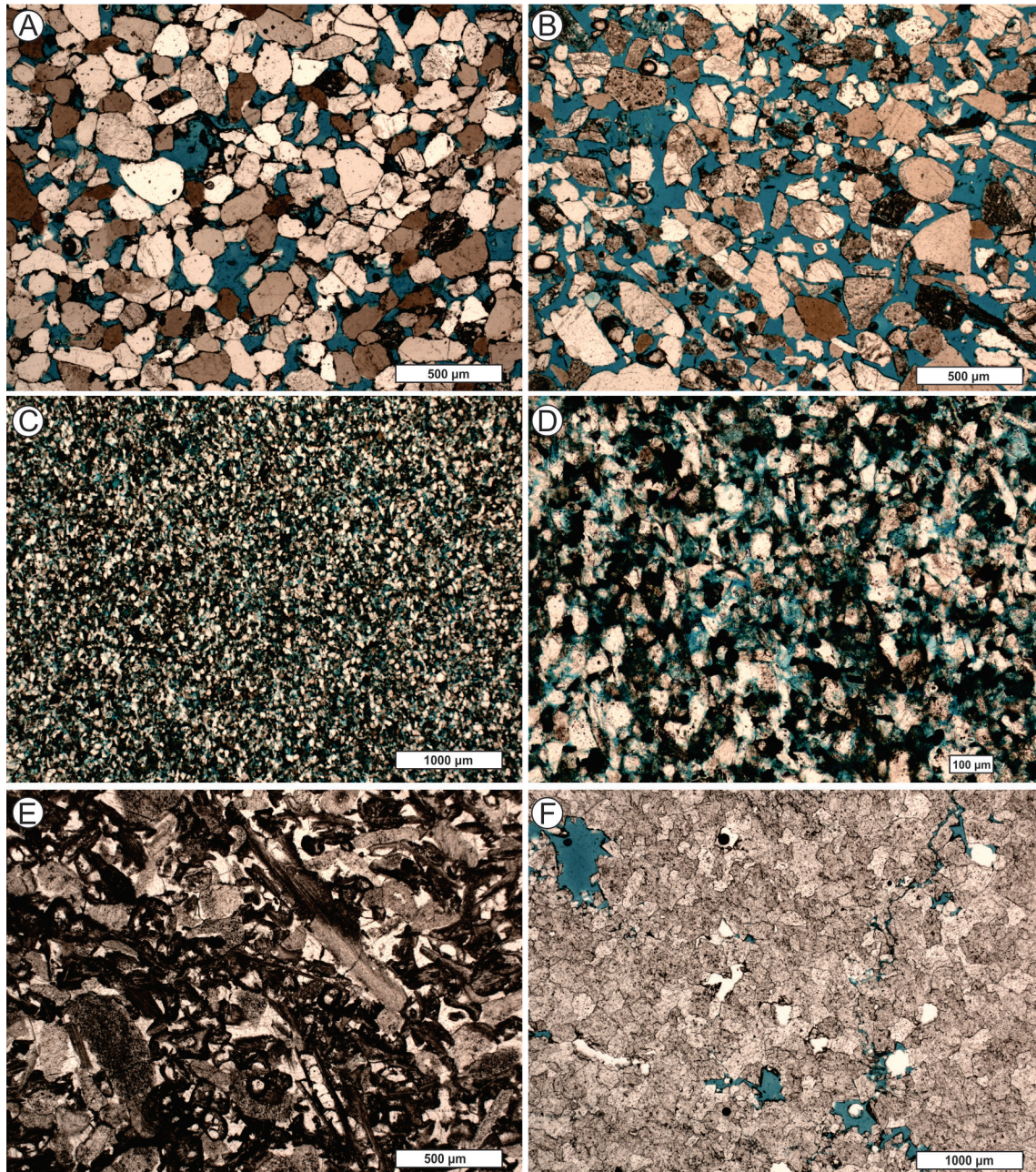


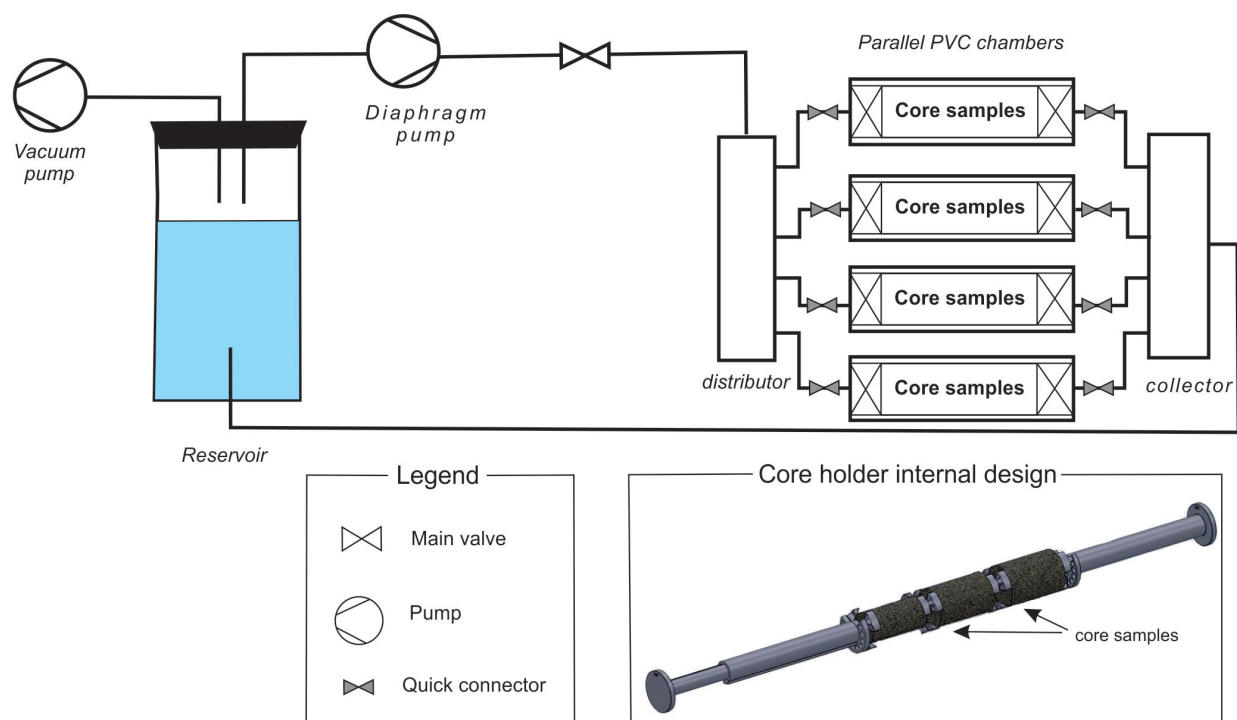


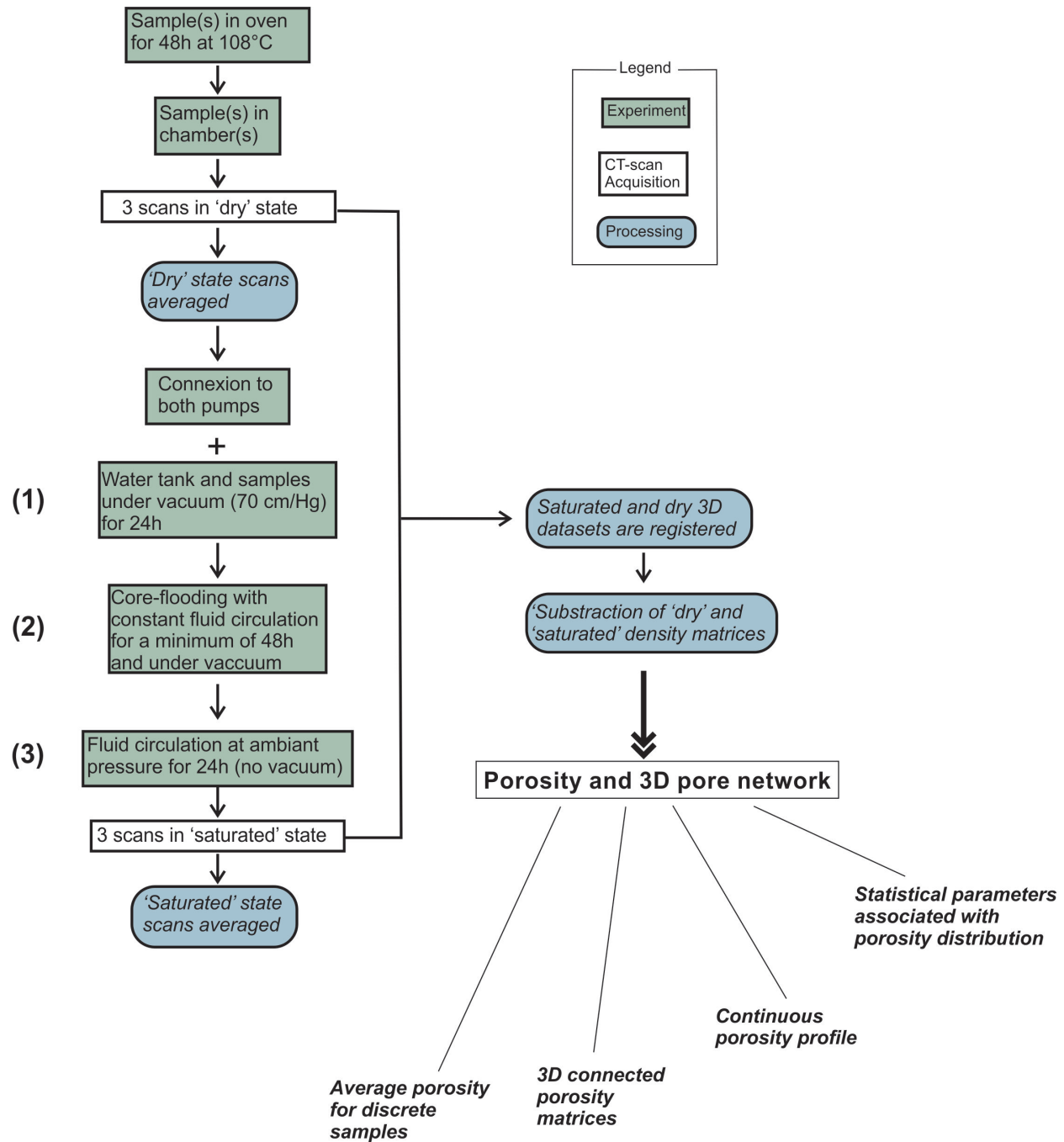


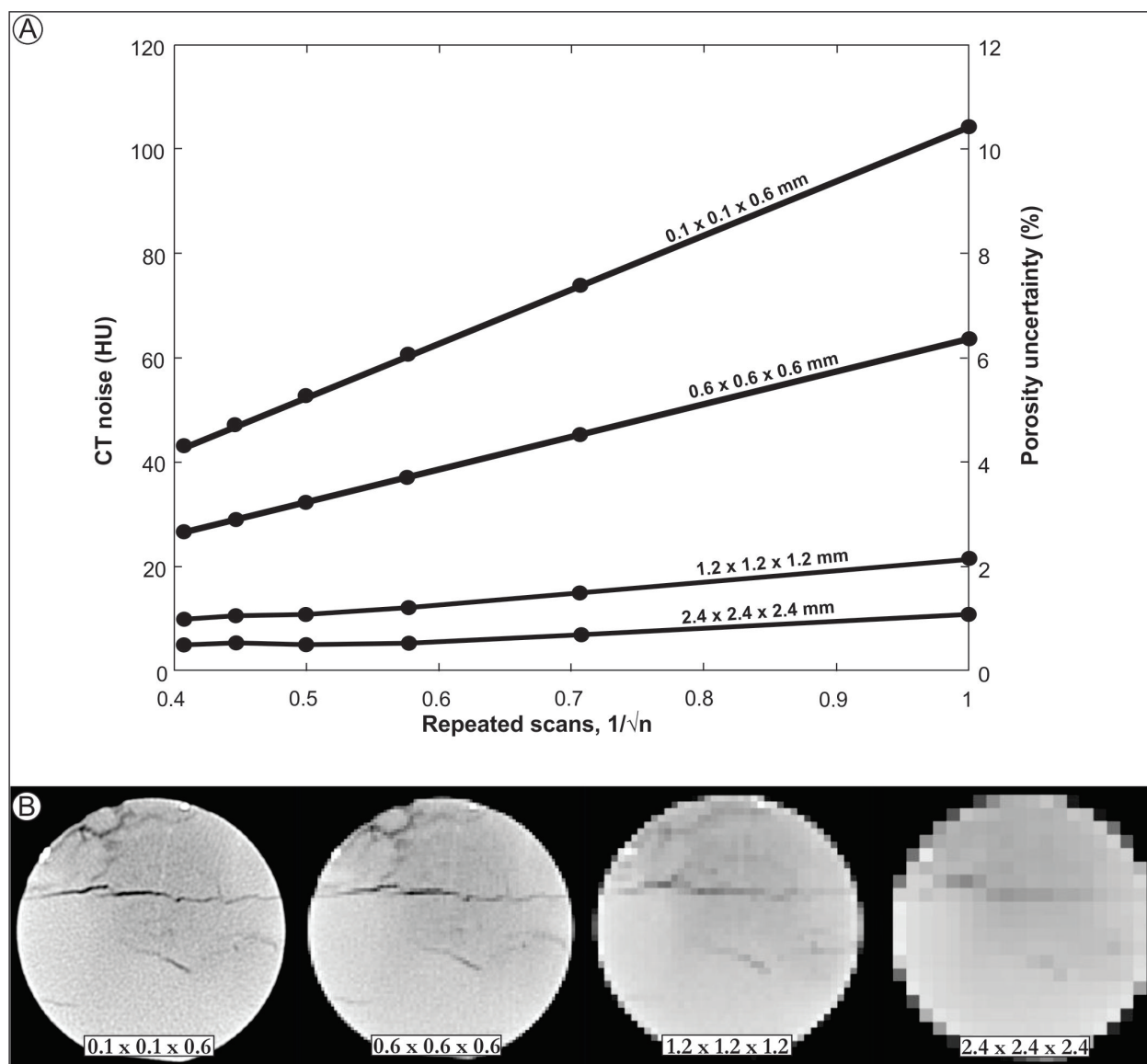


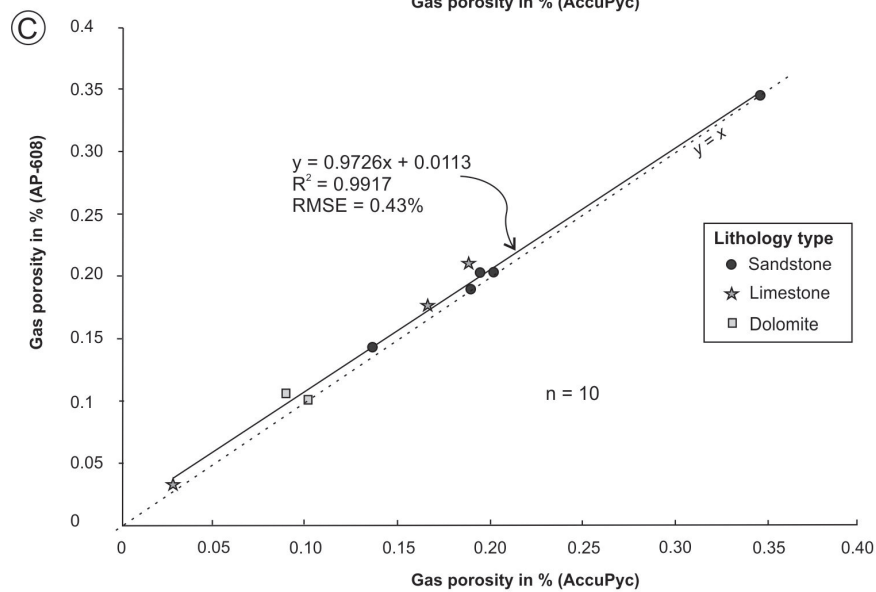
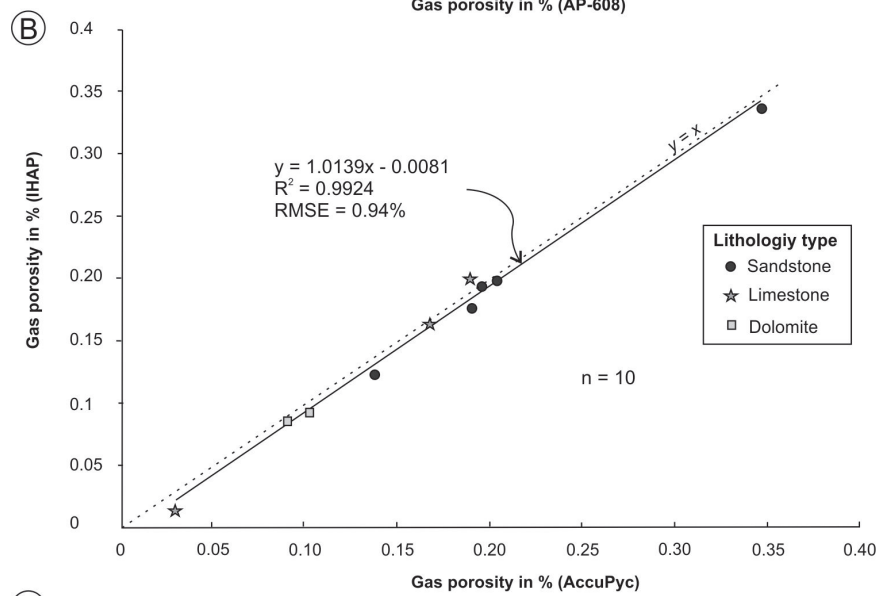
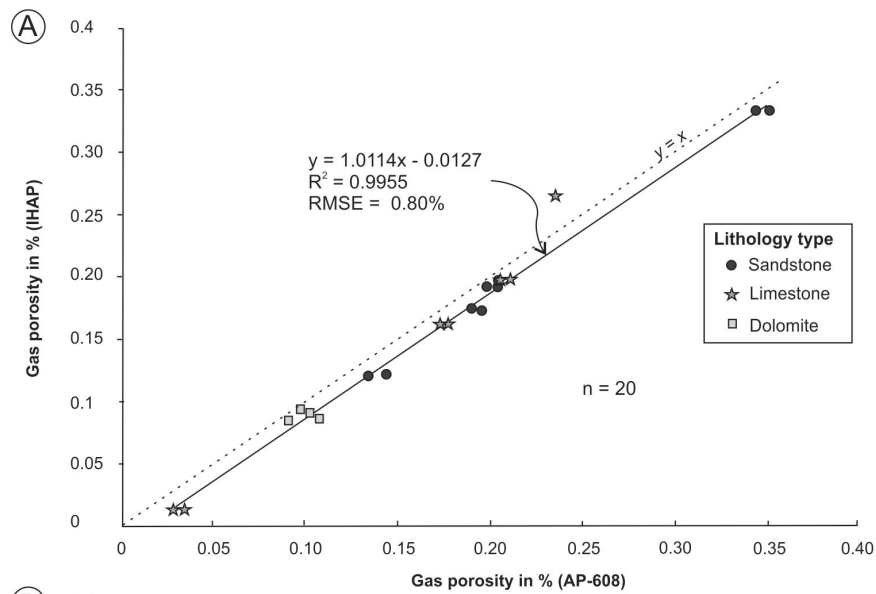


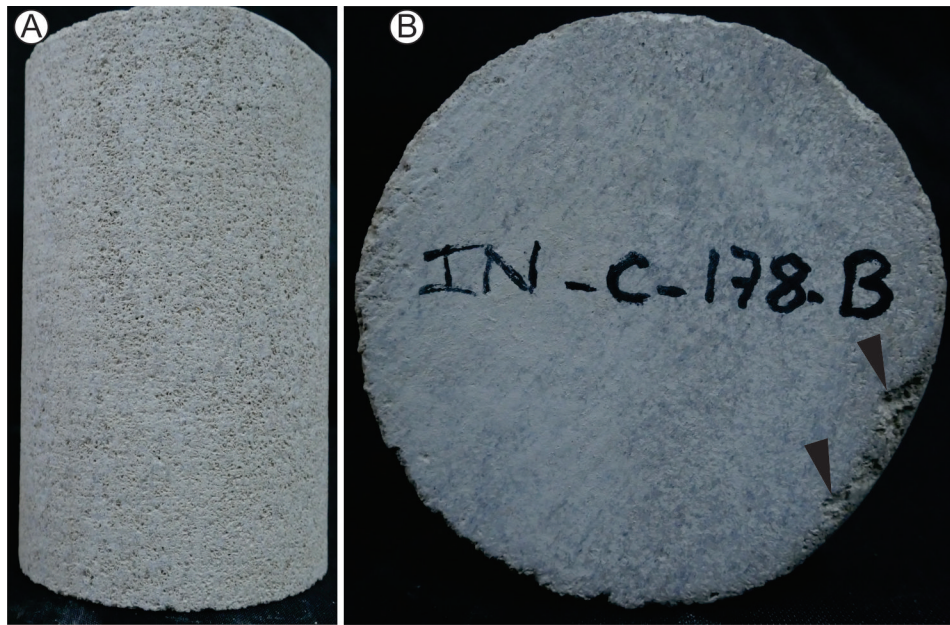


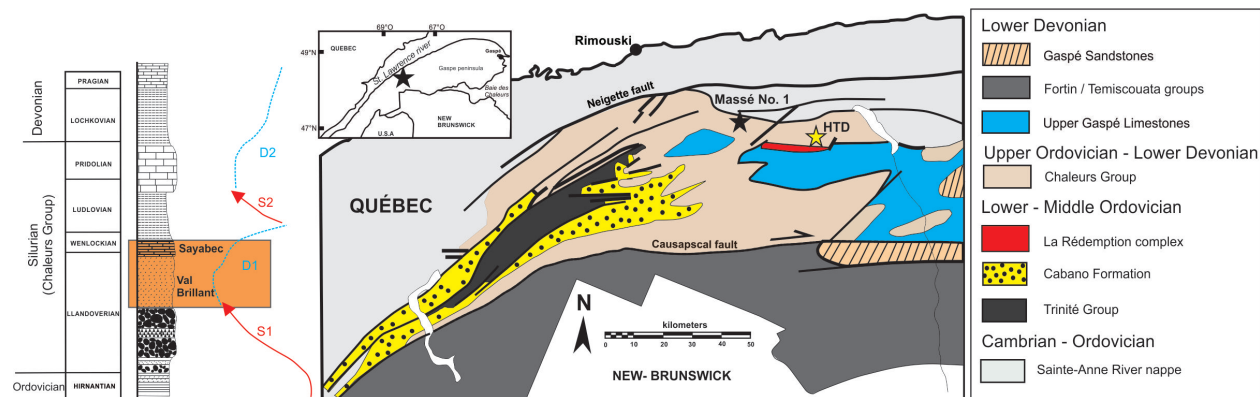


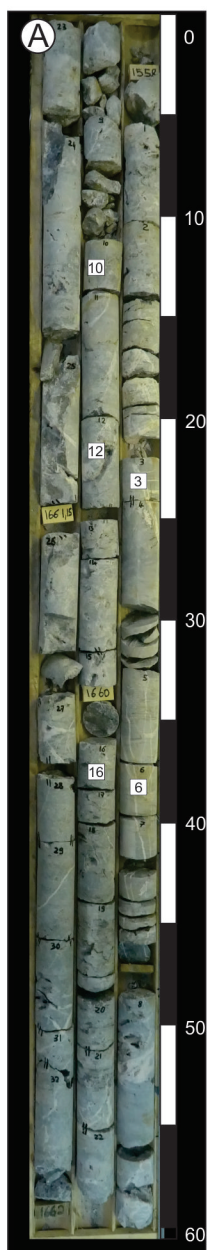












Silurian sample #3

Description

Fine-grained limestone with horizontal and oblique fractures and veins.

Porosity

Porosity was expected to be relatively low, and associated with unsealed fractures.



Silurian sample #6

Description

Fine-grained, laminated limestone with vertical and oblique fractures and veins.

Porosity

Open porosity is visible on sample surface associated with oblique fractures. However, porosity is expected to be low.



Silurian sample #10

Description

Massive dolomitized limestone with original depositional texture not visible at the macroscale.

Porosity

Thin cracks associated with visible porosity (intergranular porosity?).



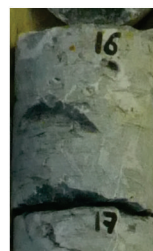
Silurian sample #12

Description

Massive dolomitized limestone largely affected by centimetric fractures and dissolution. Original depositional texture is not preserved.

Porosity

Open porosity is visible on surface and associated with large, centimetric fractures.



Silurian sample #16

Description

Bioclastic limestone with millimetric to plurimillimetric open vugs, often associated with moldic porosity.

Porosity

Open porosity is visible on surface sample as dissolved molds.

Highlights

- (1) Combined core-flooding setup and medical-CT give porosity for heterogeneous material.
- (2) Reference core material were tested and included 7 different lithologies
- (3) The new CT methodology developed strongly correlates with conventional gas porosimetry.
- (4) 3D porosity matrices and continuous porosity profiles at submillimetric scale are produced.